The Rise of E-Wallets and Buy-Now-Pay-Later: Payment Competition, Credit Expansion, and Consumer Behavior

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Abstract

The past decade has witnessed a phenomenal rise of digital wallets, and the COVID-19 pandemic further accelerated their adoption globally. Such e-wallets provide not only a conduit to external bank accounts but also internal payment options, including the ever-popular Buy-Now-Pay-Later (BNPL). We examine, for the first time, e-wallet transactions matched with merchant and consumer information from a world-leading provider based in China, with around one billion users globally and a business model that other e-wallet providers quickly converge to. We document that internal payment options, especially BNPL, dominate both online and on-site transactions. BNPL has greatly expanded credit access at the extensive margin through its adoption in two-sided payment markets. While BNPL crowds out other e-wallet payment options, it expands FinTech credit to underserved consumers. Exploiting a randomized experiment, we also find that e-wallet credit through BNPL substantially boosts consumer spending. Nevertheless, users, especially those relying on e-wallets as their sole credit source, carefully moderate borrowing when incurring interest charges. The insights likely prove informative for economies transitioning from cash-heavy to cashless societies where digital payments and FinTech credit see the largest growth and market potential.

Keywords: Cashless, Consumer Credit, FinTech, Financial Inclusion, Mobile Wallet.

JEL Codes: E42; G20; G23; G51

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1. Introduction

E-wallets, also known as digital wallets (which nest the concept of mobile wallets), contain software that enables users to make payments on their computers or mobile devices using linked bank cards, balance and savings accounts from wallet providers, or new liquidity and credit products. The past decade saw phenomenal growth in the world-wide usage of PayPal, Apple Pay, the relaunched Google Pay, and super apps such as WeChat Pay and Alipay.² The combination of digitalization and the social distancing engendered by the COVID-19 pandemic have further accelerated consumers' adoption of e-wallets for contactless payments. A conservative estimate projects the total number of e-wallet users to be over 5.2 billion globally in 2026, up from 3.4 billion in 2022.³ Together with the rise of e-wallets is the emergence of Buy-Now-Pay-Later (BNPL), a short-term FinTech credit allowing consumers to defer payments interest free into one or a few installments at the point of sale and is increasingly popular among e-wallet users. As a common offering from leading e-wallet providers around the globe, BNPL is projected to account for \$680 billion in global transactions by 2025.⁴

While economists agree that the shift from cash and bank cards toward e-wallets exerts a profound influence on the real economy (Agarwal et al., 2020b, 2022), little is known about how various payment options interact and compete within e-wallets and how BNPL affects consumer credit provision and spending behavior. On one hand, as a conduit to linked bank cards, e-wallets could reduce physical-card-carrying costs and alleviate transaction frictions and costs. Their adoption thus potentially benefits traditional financial intermediaries (e.g., banks) and consumers. On the other hand, e-wallets also offer their own payment options. For example, PayPal, a popular e-wallet in the United States, provides several internal payment options in addition to

²https://www.pwc.com/gx/en/industries/financial-services/publications/financial-services-in-2025/payments-in-202 5.html#macro3, retrieved on November 21, 2022.

³ Juniper Research. See https://www.juniperresearch.com/press/digital-wallet-users-exceed-5bn-globally-2026.

⁴ "Buy Now Pay Later Digital Spend, Led by Klarna, PayPal, and Afterpay, to Double by 2025: Reaching \$680 Billion," Kaleido Intelligence, September 2020. Alipay's Huabei, PayPal Credit, and PayPal Pay Later are examples of this fast-growing new form of credit which is typically embedded in e-wallets; Apple just added in March 2023, Apple Pay Later.

external bank-linked accounts or cards: PayPal Balance, PayPal Savings, PayPal Credit, and PayPal Pay Later. When users choose these internal payment options, e-wallets may circumvent banks, thus cutting banks off from valuable informational synergy between FinTech lending and cashless payments (Ghosh et al., 2022). This, in theory, can have an ambiguous effect on social welfare (Parlour et al., 2022). Moreover, without knowing detailed transactions and user information, it is unclear whether BNPL as a form of FinTech credit complements or substitutes bank credit, a core issue pursued in the literature (Tang, 2019; Jagtiani and Lemieux, 2019; Di Maggio and Yao, 2021), what role it plays for economies transitioning from being cash-heavy to cashless, and whether it leads to excessive spending and indebtedness (Ponce, Seira, and Zamarripa, 2017; Agarwal et al., 2020b).

We aim to bridge the knowledge gap by opening the black box of e-wallets that contain multiple payment options and conducting the first empirical investigation using unique data from a representative e-wallet provider based in China that is also a global leader in the sector. We also obtain directly matched, transaction-level data on BNPL and credit cards, instead of inferring BNPL activities from bank transactions as is previously done. The sample covers 550,000 online and 550,000 on-site transactions randomly drawn from all transactions in June 2020. We complement the e-wallet data by randomly sampling from a two-month randomized experiment conducted by the same e-wallet provider in June 2017 to study the impact of introducing BNPL credit to consumers.

Specifically, we (i) describe the distribution of payment choices in e-wallet transactions and how BNPL dominates, (ii) document how BNPL crowds out other payment options yet complements bank credit cards in payment applications, serving as a new digital cash, (iii) analyze the expansion of e-wallet credit through BNPL for consumers and regions underserved by banks, emphasizing the adoption in two-sided payment networks, and (iv) show that in the Chinese context, BNPL increases consumer spending, but does not lead to greater default or indebtedness. In addition to providing the first transaction-level analysis of e-wallets, we also join recent studies

(e.g., deHaan et al., 2022; Di Maggio, Williams, and Katz, 2022; Guttman-Kenney, Firth, and Gathergood, 2023) to add to the first canon of knowledge about BNPL and the economic implications of its meteoric emergence. Our study differs by drawing evidence from an economy transitioning from being cash-heavy to cashless and tying together the payment and credit functions in e-wallets, demonstrating that BNPL serves as a "new cash" instead of being a credit card competitor or consumer credit with negative welfare implications, which the extant literature documents in developed countries. Furthermore, we directly observe BNPL transactions with rich consumer and merchant information, circumventing any noisy inference of BNPL activities from bank transactions.

We start with summarizing stylized facts regarding the distribution of e-wallet payment options. Internal payment options—particularly BNPL—have become consumers' predominant way to pay, accounting for more than half of all transactions in our sample, whereas the most popular external option—linked debit cards—accounts for less than one-third. As the internal options dominate in e-wallet transactions, the popularity of e-wallets could disrupt traditional banks' access to payment data (Ghosh et al., 2022). Given how important payment information and digital footprints are in lending (Agarwal et al., 2020a; Berg, Fuster, and Puri, 2022), it is understandable that central banks around the world are looking into developing their own electronic payment systems (EPSs) or digital currencies to compete for payment flows with FinTech giants (Boar and Wehrli, 2021).

We next estimate a payment choice model to look into the impact of e-wallet credit expansion on other payment choices after accounting for a set of typical factors such as consumer demographics, payment preferences, merchants' acceptance, and transaction value. We find that consumer access to e-wallet credit crowds out the usage of other payment options in both on-site and online transactions. In line with the crowding-out effect, we also find that small-value transactions, which form a substantial share of payments through the e-wallet, are overwhelmingly settled with e-wallet credit through BNPL. While prior studies (e.g., Klee, 2008) conclude that physical cash is the main payment choice for settling high-frequency and small-value transactions in modern societies, we posit that BNPL is now the most popular payment choice in e-wallet transactions and hence serves as the new digital cash in transition to cashless societies.

Somewhat counterintuitively, BNPL does not compete heads-on with bank credit. Unlike prior studies that examine the relation between FinTech lending and traditional lending (e.g., Jagtiani and Lemieux, 2019; Di Maggio and Yao, 2021), our study features data including borrowers who have access to both e-wallet credit and credit cards (*dual-access* users). We can therefore rule out the potential impact of borrowers' unobservable characteristics. We find that dual-access users use e-wallet credit and credit cards in different payment scenarios, with e-wallet credit for daily, small-value transactions and credit cards for big-ticket purchases. E-wallet credit serves primarily as a potential complement of credit cards, not a substitute, because its adoption by users already with credit cards increases the spending through credit cards as well.

Because our dataset overcomes the empirical challenge in the literature concerning determining whether FinTech credit users have access to bank credit, we can investigate if e-wallet credit (i.e., BNPL) expands credit provision to those with no prior credit access. We document a substantial expansion at the extensive margin. Specifically, 84.14% and 78.38% of on-site and online consumers without credit cards, respectively, can now access e-wallet credit. Moreover, 43.54% of on-site merchants who previously did not accept credit cards now accept e-wallet credit. Additionally, 44.37% of on-site and 90.21% of online transactions that could not be completed using credit cards can now be completed using e-wallet credit. The combination of demand from consumers and merchants for BNPL, and the supply by the e-wallet provider, has led to the exponential growth of BNPL usage. The impact of BNPL access on consumer credit usage also appears more pronounced in less-developed areas such as rural areas and northern regions and for women. Overall, the e-wallet credit seems to benefit those disadvantaged or underserved by banks.

With the wide use of BNPL, a potential concern is that the ease and convenience of

this FinTech-based consumer credit may induce consumers, especially unsophisticated ones who lack financial literacy and budgeting education, to overborrow and overspend (Berg, Fuster, and Puri, 2022; Bu et al., 2022). We exploit a two-month randomized experiment conducted by the e-wallet provider in June 2017 to examine the impact of the introduction of BNPL on consumer spending. For users who were extended credit in this experiment, we first document a significant consumption-boosting effect consistent with that observed in developed economies such as the United States: an increase equivalent to 4.78% of the average monthly consumption as reported by the National Bureau of Statistics of China.

Next, we return to our transaction data and analyze how consumers change their usage of BNPL when they have unpaid debt and incur interest expenses. We distinguish between two types of users according to their credit access: *single-access* users (e-wallet credit users with no linked credit card), and *dual-access* users (mentioned earlier). While single-access users appear to use e-wallet credit more frequently, they use the credit cautiously, as the proportion of revolvers (those who incur interest charges on unpaid debt) and the unpaid-debt ratio for revolvers are lower. They also reduce BNPL usage in transactions once they incur interest expenses from late payments (which are higher than the rates from credit cards), likely as an attempt to improve their credit score and to pave a pathway to bank credit access (Agarwal et al., 2021). To some extent, this finding alleviates our concern that users who receive consumer credit for the first time overuse e-wallet credit due to the lack of financial literacy and budgeting education.

Our findings have broad implications. First, given that payment networks have constituted the core products of BigTech giants and FinTech startups, our investigation on the distribution of payment choices in e-wallet transactions provides an initial benchmark for understanding their role in large digital ecosystems. Second, because e-wallet providers have high-frequency and exclusive data on merchants and consumers underserved by banks, discussions on "open-FinTech"—in addition to the "open banking" proposal—can be timely. Third, although BNPL raises concerns about

consumer indebtedness worldwide, our study suggests that BNPL users, especially those with no credit access from banks, carefully moderate credit usage in an economy transitioning from cash-heavy to cashless. Credit expansion through BNPL does not necessarily come at the expense of greater consumer indebtedness or default under inclusive, ex-ante screening and ex-post monitoring and incentive provision of the e-wallet provider.

Our data and the Chinese setting have several appealing features well-suited for the economic questions we study. First, China is the largest e-wallet market in the world and currently has the largest number of BNPL users (Section 2.1 contains more details), with mature mobile payment networks.⁵ Second, the e-wallet functionality and BNPL design in our study are reasonably representative, not only in China with its around one billion users but also on the world stage. For example, the BNPL embedded in the e-wallet we study offers a revolving credit line and allows installment payments, thus combining the features of PayPal Credit and PayPal Pay Later, and is similar to other major BNPL providers such as Affirm in terms of "Pay-in-4" (installments) products and credit lines. Both the representativeness in terms of functionality and the fact that many e-wallet providers around the globe are converging to the Chinese model (CFPB, 2022 and Section 2.1) enable our findings to be useful for other e-wallet and BNPL providers, even though many of them started as pureplay payment or credit products (thanks to the low-interest environments). Finally, our dataset is likely the only one in the literature that allows direct observations of transactions in e-wallets and involving BNPL, covering both online and on-site transactions and both merchant and consumer information, thanks to that our e-wallet provider has its own e-commerce platforms and QR code registration system.

Our work relates to several strands of literature. FinTech firms typically start with payments and then expand into lending, insurance, investment products, etc. (Frost et

⁵ The mobile payment boomed early in 2011 and the adoption among Chinese adults reached 82% in 2017. The coverage and usage had since stabilized. See https://idf.pku.edu.cn/docs/20210421101507614920.pdf and http://www.pbc.gov.cn/goutongjiaoliu/113456/113469/3602384/index.html; accessed on August 6, 2022.

al., 2019). However, the existing literature appears to focus predominately on lending, largely due to data limitations on payment transactions. As highlighted by Berg, Fuster, and Puri (2022), payment data are fundamental for credit ratings in the lending business. Our study focuses on FinTech payments and provides the first empirical description on a timely and important issue: payment choices within e-wallets.⁶ Although there have been payment surveys in advanced economies and numerous studies on (electronic) payment choices (e.g., Arango et al., 2015; Koulayev et al., 2016; Wang et al., 2016; Crouzet, Gupta, and Mezzanotti, 2019; Agarwal et al., 2020b), these surveys have always categorized e-wallets, if at all, as an aggregate category. Looking into the black box of e-wallet payment options is important for understanding current changes in payment systems and evaluating e-wallets' economic impacts. Our analysis of both the consumer access to and merchant adoption of BNPL as a payment option also adds to studies such as Higgins (2019) that analyze the coordinated technology adoption in two-sided markets.

We contribute to the literature on FinTech credit, especially the emergent discussions on BNPL. Earlier studies focus on credit for online merchants, small firms, and entrepreneurs (e.g., Huang et al., 2018; Hau et al., 2019; Hau et al., 2021, Gopal and Schnabl, 2022), whereas we study credit for consumers and households. Within this domain, marketplace lending has been shrinking while BNPL has been growing. Berg, Fuster, and Puri (2022) estimate that new lending in the U.S. marketplace lending market was \$6 billion in 2020; the U.S. BNPL market, by contrast, was about \$25 billion. Compared to the numerous studies on marketplace lending (e.g., Lin, Prabhala, and Viswanathan, 2013; Freedman and Jin, 2017; Jagtiani and Lemieux, 2019; Tang, 2019; Cong, Tang, Xie, and Miao, 2019), the research on consumer FinTech credit, particularly BNPL, is scarce.

Several contemporaneous articles study BNPL in developed economies where

⁶ Hong, Lu, and Pan (2020) document that repeated usage of digital payments through super-apps can help users build familiarity and trust, thereby increasing their participation and risk-taking in mutual fund investments offered by super-apps.

credit cards have already largely replaced cash. Guttman-Kenney, Firth, and Gathergood (2023) describe BNPL usage in the UK, where 19.5% of active credit card users had a transaction with a BNPL firm on their credit cards in 2021, and young users and those living in deprived regions with limited repayment capability had an even higher ratio. deHaan et al. (2022) find that BNPL in the United States plausibly has a negative effect on users' spending habits and financial health. Similarly, Di Maggio, Williams, and Katz (2022) find that BNPL access increases total spending and the proportion of retail spending in the United States, which is better explained by the "liquidity flypaper effect." These studies provide valuable information, especially time trends, on BNPL, but have to rely on bank and credit card transactions to infer BNPL activities, thus capture only transactions linked to banks, with limited matched merchant or consumer information and primarily cover a small set of pure-play providers.⁷ A more recent study by Berg et al. (2023) studies another BNPL model offered by a furniture merchant in Germany to show, through a randomized controlled trial, that BNPL induces greater consumer spending and benefits the merchant more than PayPal does. The authors focus on the cost-benefit analysis of a particular type of "pay later" that must occur within 14 days and is typically covering larger transaction amounts which are different from typical BNPL the rest of the studies analyze.

We investigate the usage of BNPL and its impact on consumer spending and indebtedness directly using a comprehensive, matched, transaction-level dataset, in a country with drastically different financial development and business ecosystems, while featuring an equally important BNPL market and business model that many markets around the globe are converging to. We highlight the payment nature of BNPL and examine its interaction with other payment options. Furthermore, our data are representative of the population of BNPL users from both sides of the two-sided payment markets because they are not limited to one merchant in a particular product category or merchants working with banks.

⁷ For example, PayPal is excluded because it contains large non-BNPL business, despite also being a major BNPL player. The transactions from omitted BNPL providers are labeled as non-BNPL, leading to classification errors.

Finally, we complement earlier research on mobile payment, consumer credit, and overspending (Aydın, 2022; Gross and Souleles, 2002; Ponce, Seira, and Zamarripa, 2017; Agarwal et al., 2020b). Previous studies document a spending-boosting effect of payment innovations (e.g., mobile payment and e-wallets), which causes concerns about overspending, consumer addiction, and debt traps that have plagued many developed economies (e.g., Manning, 2001; Williams, 2004). Recently, Wang (2023) theoretically and empirically analyzes competition among bank-card-based networks and finds that it leads to overuse of credit and reduced social welfare. We find BNPL, as a new consumer credit, boosts consumption. But it does not increase household indebtedness in China, a phenomenon also documented in the Israeli market (Gelman and Roussanov, 2022), suggesting that the payment competition induced by e-wallets and BNPL in China is rather nuanced: Heterogeneity in payment options leads to different merchant fees, retail prices, and consumer segments, with BNPL serving many consumers without credit card access. Consequently, depending on the payment design and market environment, merchant fees do not necessarily get passed onto retail prices, and consumers do not necessarily overuse the newly available credit.

The rest of this paper is organized as follows. Section 2 provides the institutional background and introduces the data. Section 3 describes the landscape and stylized patterns of e-wallet payments, including the effects of BNPL on other payment options and bank consumer credit. Section 4 presents the findings concerning e-wallet credit expansion through BNPL. Section 5 investigates how the rise of BNPL affects consumer spending and indebtedness. Section 6 concludes.

2. Institutional Details and Data Description

2.1 Institutional Background

The emergence of the e-wallet economy. E-wallets allow users to store digital currency and use it to make online transactions as well as on-site transactions via scanning QR codes. E-wallets are steadily replacing the use of physical cards and cash,

and are hailed as revolutionizing digital finance, transforming how consumers and markets operate, empowering businesses, and boosting financial inclusion. For one, they significantly reduce the inconvenience of carrying physical cards and cash and are faster and easier to use relative to bank-based payments and transfers. Many consumers are purchasing products through e-commerce platforms with various e-wallets, not to mention that digital currency users are also using e-wallets for secure transactions.

Digital wallets first became popular in developing countries that were considered cash heavy or lacking well-developed credit reference systems.⁸ After they took off in China and India, they were quickly adopted in countries such as Philippines, Thailand, and Vietnam. Australia and Singapore also embraced e-wallets quickly, followed by other developed economies in Europe and North America amidst the COVID-19 pandemic because they offer a convenient payment option without physical contact.

Depending on the issuer and the level of acceptance, e-wallets come in three types: closed e-wallets issued by a specific merchant or service provider (e.g., Amazon Pay), semi-closed e-wallets that allow users to make purchases at multiple merchants but lack widespread applicability (e.g., Alipay and Paytm), and open e-wallets that are issued by banks or institutions partnered with banks and allow users to make purchases at any merchant that accepts electronic payments (e.g., PayPal, Apple Pay, and Google Pay).

BNPL around the globe. BNPL, a new FinTech consumer credit, has also emerged rapidly together with e-wallets over the past decade and especially during the pandemic. While there is no single definition of BNPL, it is generally defined as a type of short-term consumer credit that allows users to make purchases but defer payments into one or a few instalments. With the pandemic limiting household income worldwide and banks cautious from providing credit in challenging economic

⁸ While the technologies behind digital wallets seem new, some have been around for decades. For example, Coca-Cola's vending machines back in 1997 allowed people to pay via text message; Amazon launched 1-click in 1999 and PayPal expedited eBay purchases in 2022.

environments, BNPL meets the elevated demand for non-contact payments and liquidity. As a result, established e-wallet providers such as Alipay and PayPal, as well as FinTech start-ups such as Affirm and Klarna, have all incorporated BNPL as a core product.

BNPL providers implement two distinct strategies to acquire customers: one is the merchant partner model that cooperates with retailers to embed BNPL products on the merchants' checkout page; the other is the app-driven model that offers BNPL products either through e-wallets or "pure-play" BNPL apps and allows users to make purchases under a predetermined credit limit (Federal Reserve Bank of Kansas City, 2021; CFPB, 2022). Like credit cards, BNPL provides a line of credit to consumers for online or on-site purchases, but with lower fees and less stringent credit history requirements for merchants and consumers. In general, BNPL products are interest-free, but they may carry interest for late payments or loans with longer terms. For example, a typical BNPL product provides the option to pay in four installments, with no interest charged. At checkout, consumers make a down payment of 25 percent, and the remaining three installments are due in two-week intervals. In some cases of large, infrequent purchases with loan term lengths up to three years, a Point-of-Sale (POS) installment is also categorized as BNPL.9 For POS installments, down payments are typically not required, but monthly payments include interest or fixed finance charges. Importantly, and different from traditional payment installments offered by merchants and stores, BNPL providers or the associated e-wallet providers supply the BNPL credit, not the merchant or store.

The global BNPL market size is U.S. \$103.6 billion in 2022 and is projected to reach U.S. \$467.34 billion in 2026, and even U.S. \$3.98 trillion by 2030.¹⁰ Among the top 20 BNPL nations ranked by the proportion of BNPL transactions in

⁹ A report by McKinsey mentioned POS installment as a type of BNPL lending. See https:// www.mckinsey.com/industries/financial-services/our-insights/buy-now-pay-later-five-business-models-to-compete ¹⁰ Research and Markets report on global BNPL market. https://www.researchandmarkets.com/reports/5700456/ buy-now-pay-later-global-market-report-2022?utm_source=BW&utm_medium=PressRelease&utm_code=tgbvg4 &utm_campaign=1793416+-+Buy+Now+Pay+Later+Global+Market+Report+2022%3a+Increase+in+the+Adopti on+of+Online+Payment+Methods+Drives+Growth&utm_exec=como322prd.

e-commerce market, 12 nations are from the Europe, 7 nations are from the Asia-Pacific regions, and one nation is from North America.¹¹ In the United States, the five leading BNPL providers originated BNPL loans totaling U.S. \$24.2 billion in 2021, with an annualized growth rate of over 200 percent since 2019 (CFPB, 2022).¹² Australia, home to one of the earliest and most successful BNPL provider Afterpay, features more than 6.1 million BNPL users as of June 2019, representing up to 30% of the adult population; the BNPL transaction value increased from \$3.1 billion to \$5.6 billion in 2017-2019, with the growth rate up to 79 percent (ASIC, 2020). In Europe, BNPL accounted for 11.7% of e-commerce share in Denmark, 12.8% in Finland, 18.1% in Norway, 25.2% in Sweden, and 19.7% in Germany, as of the end of 2021.¹³ Asia also records strong BNPL growth. According to the report released by Mordor Intelligence, the transaction value of BNPL in India had reached about U.S. \$7 billion in 2022, with 22% of Indian consumers using BNPL to make purchases.¹⁴

E-wallets and BNPL in Asia and China. Asia has led the world in e-wallet economy. In 2021, e-wallets accounted for 69% of e-commerce spending and 44% of point-of-sale transactions in Asia-Pacific regions.¹⁵ The underdeveloped payment infrastructure, a high proportion of unbanked population, and fast-growing e-commerce markets contribute to the emergence of e-wallet super apps such as Alipay and WeChat Pay in China, Grab in Singapore, Gojek in Indonesia, Kakao in South Korea, and Paytm, PhonePe, and Vodafone M-Pesa in India. These e-wallet providers started from scratch and attracted huge consumer base underserved by banks via apps and QR codes. They provide a range of services, including payments, credit, wealth management, and everyday life services. The frequent use of the payment function helps customers build familiarity and trust with e-wallet super apps, which facilitates the use of other services (Hong, Lu, and Pan, 2020). As *Financial*

 ¹¹ https://www.paymentscardsandmobile.com/top-20-bnpl-nations-ranked-by-e-commerce-size-and-market-share/.
 ¹² The five BNPL providers include Affirm, Afterpay, Klarna, PayPal, and Zip (formerly Quadpay).

¹³ Worldpay from FIS, the global payments report for financial institutions and merchants. Accessed from https://offers.worldpayglobal.com/rs/850-JOA-856/images/ENGPR2022.pdf.

¹⁴ Mordor Intelligence, India buy now pay later services market-growth, trends, COVID-19 impact, and forecasts (2023-2028). https://www.mordorintelligence.com/industry-reports/india-buy-now-pay-later-services-market.

¹⁵ Worldpay from FIS, the global payments report for financial institutions and merchants. Accessed from https://offers.worldpayglobal.com/rs/850-JOA-856/images/ENGPR2022.pdf.

Times puts it, the "e-wallet economy" in Asia is a digital financial revolution that has boosted financial inclusion and empowered business by bringing billions of users into a quasi-banking ecosystem.¹⁶

Among Asian countries, China has the most developed e-wallet market, with e-wallets being the most popular payment method and accounting for 83% of e-commerce transactions in 2021. The corresponding number in India, Indonesia, and Philippines are 45.4%, 38.8%, and 30.5%, respectively.¹⁷ Alipay and WeChat Pay are the two largest e-wallets in China, serving 1.3 billion and 0.9 billion users as of 2022 respectively.¹⁸ The data for our study come from one of these leading e-wallet providers in China. The allowable credit amounts for BNPL through these digital wallets, like credit cards, are based on some form of credit reference systems, which in turn builds on past activities within the ecosystem such as historical transactions.

In contrast, e-wallets initially follow different trajectories in North America and Europe, where bank cards acceptance are nearly universal, and cards still dominate the payment systems. In 2021, e-wallet transactions accounted for only 10% point-of-sale transactions in North America and 7.7% in Europe.¹⁹ Several prominent e-wallet providers in the United States, including PayPal, Apple Pay, and Google Pay, used to create a digital interface for physical cards and mainly relied on existing advanced card payment networks to process digital payments. However, as the COVID-19 pandemic has greatly increased the demand for e-wallets, these e-wallet providers are also trying to incorporate more functions such as wealth management and racing to create "super apps" mimicking those in Asia.²⁰

Finally, although the United States has the largest BNPL transaction volume,

¹⁶ Financial Times, "Wallet Wisdom: The Transformative Power of Asia's "Wallet Economy." See https:// www.ft.com/partnercontent/mastercard/wallet-wisdom-the-transformative-power-of-asias-wallet-economy.html

¹⁷ Worldpay from FIS, the global payments report for financial institutions and merchants. Accessed from https://offers.worldpayglobal.com/rs/850-JOA-856/images/ENGPR2022.pdf.

¹⁸ Data source: https://merchantmachine.co.uk/the-countries-most-reliant-on-cash-in-2022/.

¹⁹ Worldpay from FIS, the global payments report for financial institutions and merchants. Accessed from https://offers.worldpayglobal.com/rs/850-JOA-856/images/ENGPR2022.pdf.

²⁰ PaymentsDive: "Super-apps" are the next evolutionary step for financial services. Accessed from https://www.paymentsdive.com/news/super-apps-are-the-next-evolutionary-step-for-financial-services/601424/.

China is currently the largest BNPL market with over 500 million users as of 2020.²¹ The BNPL market in China was about U.S. \$30 billion in 2020 and is projected to reach U.S. \$750 billion in 2025.22 While BigTechs such as Ant Group, JD, and Pinduoduo are dominating China's BNPL market, global players including Afterpay and Klarna have entered the market through strategic investment or establishment of subsidiaries. Unlike Australia and Europe where BNPL was usually launched by FinTech startups, a prominent feature in China is that BNPL is launched by leading e-wallets with internal e-commerce systems, such as Huabei in Taobao and Tmall and Baitiao in JD e-commerce platforms. This feature also manifests itself in a major BNPL provider in the United States, PayPal, which incorporates BNPL into its well-established e-wallet payment system. Notably, against the increasing interest rate environment worldwide, the marginal profit of BNPL providers is decreasing, making it more important to rely on a larger user base and cross-selling capacities to ensure sustainability. Consequently, an increasing number of major BNPL providers around the globe are creating their proprietary super-apps and offer BNPL within self-contained e-commerce ecosystem, converging to China's business template for BNPL practices (CFPB, 2022).

2.2 Data

E-wallet data. Our first dataset contains over one million e-wallet transactions from a leading e-wallet provider in China. Supplementing this dataset are consumer- and merchant-level statistics that contain information on the two-sided market: buyers' demographic information and sellers' basic information including sales and industries. This e-wallet was initially launched on one of the world's largest e-commerce platforms as a payment tool to solve the problem of trust between buyers and sellers in online transactions, before growing into a multi-functional app with around one billion users. Its users are required to be at least 16 years of age and have

²¹ Accessed from https://www1.hkexnews.hk/listedco/listconews/sehk/2020/1026/2020102600165.pdf.

²² Forbes, "Buy Now Pay Later in Asia: The Drivers and Issues." Accessed from https://www.forbes.com/sites/zennonkapron/2022/11/02/buy-now-pay-later-in-asia-the-drivers-and-issues/.

less than 1.5% of users aged above 60. The total number of domestic users as of June 2019 is about the same as (slightly bigger than) the size of the population aged 16-59 in China, based on the 2018 census.

The unit of observation is checkout transaction, which represents one consumer's total purchase from one merchant at the point of sale. The data sample is comprised of 550,000 on-site transactions and 550,000 online transactions randomly drawn from all checkout transactions through this e-wallet provider in June 2020, whose distribution of payment options and purchase categories most closely mimics annual averages in 2019, 2020, and 2021, and is free from the effects from national holidays.²³ We observe information commonly found on most receipts from a purchase online or in a grocery store: the merchant identification number, consumer identification number, date and time of the transaction, transaction value, and payment instrument.

In a typical transaction, consumers face five payment options: e-wallet balance, e-wallet savings, e-wallet credit through BNPL, linked debit cards, and linked credit cards.²⁴ We exclude transactions with values lower than RMB ¥1 (approximately US\$0.15), which are mostly from click farming and are a small fraction of all transactions, to avoid biasing our estimation. Furthermore, in the uncommon case of transactions combining multiple payment options, we record the payment option with the highest associated transaction value. Note that when merchants decide on whether to accept an e-wallet, they automatically accept the linked debit card, e-wallet savings, and e-wallet balance options. However, they have choices on whether to accept the linked credit card and/or BNPL payment options.

The data provider internally matches consumers' and merchants' information to each transaction. Our data therefore contain information on the merchants, including (1) *average monthly sales*, the average value of monthly sales during the

²³ The e-wallet allows peer-to-peer transfers on a limited scale, which is not included as payment transactions because they are much smaller in scale and the data is not accessible due to the Data Security Law and the Personal Information Protection Law introduced in June and August of 2021.

²⁴ Coupons are associated only with e-commerce or on-site store promotion programs and often have to be combined with other payment options. Though in 2018 and 2019, some coupons were offered to first-time BNPL users, they are negligible in our 2020 data, and most coupons can be combined with any payment option.

last twelve months; (2) *D_Merchant_CC*, a dummy variable that takes value of 1 if the merchant accepts payment by credit card; (3) *D_Merchant_BNPL*, a dummy variable that takes value of 1 if the merchant accepts payment through BNPL; and (4) *Industry*, the industry that the merchant belongs to.

Concerning consumers, our data cover (1) D_Consumer_CC, a dummy that takes value of 1 if the consumer has linked credit cards; (2) D_Consumer_BNPL, a dummy that takes value of 1 if the consumer has access to e-wallet credit through BNPL; (3) *Female*, a dummy variable that takes value of 1 for females and zero for males; (4) Age, the age of the consumer; (5) e-wallet average monthly flow in the last year, a measure that proxies for consumers' activeness in using the e-wallet and also (coarsely) for consumers' financial position in the sense of "cash flow"; and (6) *e-wallet wealth*, the total amount of savings and other wealth products of a user in the e-wallet. Similar to the variable *e-wallet average monthly flow in the last year*, the variable e-wallet wealth proxies for consumers' activeness in using the e-wallet and also (roughly) for consumers' financial position in the sense of "balance." We also create variables capturing consumers' payment preferences: how long since each payment option was adopted by a consumer (Option adoption length), and each consumer's most preferred option, which is set as the default payment option (D_preferred_option). They are useful in controlling for behavioral or cognitive biases, e.g., the default payment option gets used while a user would have preferred another option in its absence.

 Table 1. Summary statistics of the main variables

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| Variables | Ν | Mean | Std | P25 | P50 | P75 |
|--|-----------|-------|------|------|-------|-------|
| A. Transaction and merchant levels | | | | | | |
| Transaction value (log) | 1,079,193 | 3.40 | 1.32 | 2.40 | 3.32 | 4.26 |
| Average monthly sales (log, in 10,000 RMB) | 1,076,689 | 12.37 | 3.97 | 9.22 | 11.88 | 14.85 |
| D_Merchant_CC ^d | 1,087,826 | 0.63 | 0.48 | 0 | 1 | 1 |
| D_Merchant_BNPL ^d | 1,100,000 | 0.75 | 0.43 | 1 | 1 | 1 |
| B. Consumer level | | | | | | |
| D_Consumer_BNPL ^d | 1,095,838 | 0.85 | 0.36 | 1 | 1 | 1 |
| D_Consumer_CC ^d | 1,100,000 | 0.31 | 0.46 | 0 | 0 | 1 |
| Female ^d | 1,091,673 | 0.53 | 0.50 | 0 | 1 | 1 |
| Age | 1,071,705 | 32.74 | 9.58 | 25 | 31 | 38 |
| E-wallet average monthly flow in last year (log) | 1,086,206 | 9.71 | 2.98 | 8.70 | 10.42 | 11.61 |
| E-wallet wealth (log) | 1,090,573 | 4.47 | 3.58 | 0.43 | 4.63 | 7.30 |

Note: The data are winsorized at the 0.5% level. Superscript d indicates the dummy variables.

Table 1 presents summary statistics of the main characteristics of merchants and consumers, weighted by transaction frequencies as in Klee (2008). The sample exhibit significant variations in the demographics, with merchant age ranging from less than a year to over five years, and merchant sales varying substantially, implying that the transactions represent the sales from a wide range of products and services. For consumers, although there is significant variation in e-wallet flow and savings, their age is concentrated between 20 and 40 years old, consistent with the overall age distribution of e-wallet users. Finally, merchants' acceptance of and consumers' access to e-wallet credit and credit cards vary substantially.

Data from a randomized experiment. Furthermore, we augment the transaction dataset by incorporating a second dataset derived from an earlier BNPL experiment conducted by the consumer credit department from the same e-wallet provider in June 2017 on millions of users who already met the credit eligibility criteria but had not yet been offered BNPL. Eligibility was determined mainly based on users' payment and consumption histories, which indicated how much consumer credit the users could and would repay. The requirement was much less stringent than that for credit cards, as evidenced by the less than 5% of our randomized sample of eligible users from the experiment having credit card access, and by the fact that more than one third of the Chinese population had utilized BNPL by 2020. Users residing in

Beijing, Shanghai, Guangzhou, and Shenzhen were excluded because of promotional BNPL activities in these large cities prior to the experiment.

On June 1, 2017, the company randomly assigned users treatment and control. The treated individuals were offered BNPL and were notified of their new credit limits immediately. The individuals in the control group had no access to BNPL and did not anticipate such access in the near future. The experiment ended after two months, in August 2017, when the control group was granted BNPL as well. The e-wallet provider did not notify its users that they were part of an experiment, which eliminated the possibility of treatment group members' changing their behavior in response or anticipation of participation or BNPL usage in any near term.

Our second dataset is a subsample of users in this experiment, including 700,000 user-month observations for 100,000 users between January and July 2017.²⁵ Specifically, the company randomly drew 50,000 users from the original treatment group and matched 50,000 users from the original control group to treated users based on their consumption and payment histories. Since we do not observe the separate trends for the level of consumption for treated and untreated users, we display the ratio of consumption of treated to untreated users to check for parallel pre-trends (see Figure 1). The ratio was between 0.99 and 1.00 before the experiment, indicating no significant difference in the levels and trends of monthly consumption between the two groups in each month prior to the experiment. We exploit this experiment mainly to study the effect of BNPL access on consumption.

²⁵ Due to the Data Security Law and the Personal Information Protection Law, we are not allowed to access the full sample. Exported variables are typically restricted to unitless ratios.

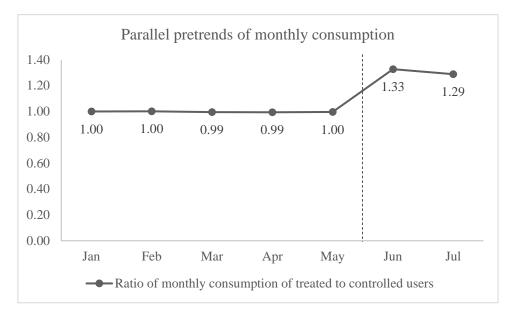


Figure 1. Parallel pre-trends of consumption in the pretreatment period

3. Digital Payment Competition Amidst the Rise of BNPL

Our data allow us to examine the competitive landscape of payment options, with a particular focus on the rise of BNPL. In particular, we investigate how the expansion of BNPL access affects payment method choices, and how BNPL interacts with credit cards, both of which offer short-term consumer credit in addition to being a payment option. We find that expansion of BNPL access has significantly displaced other payment methods, but it does not serve as a direct substitute for credit cards. BNPL appears to complement bank credit even for dual-access users—those with both credit cards and BNPL as directly observed in our data—and serve as the new digital cash in transaction scenarios different from credit card transactions.

3.1 Opening the E-Wallet Blackbox: The Distribution of Payment Types

Our data enable us to offer the first empirical description and usage distribution of prominent payment options in e-wallets: linked debit cards, linked credit cards, e-wallet balance (digital cash stored in e-wallet accounts), e-wallet savings (wealth-management products provided by e-wallet providers that function as interest-bearing demand deposits), and e-wallet credit through the FinTech credit product, BNPL.

How users pay dictates how money and information flow through the plumbing of the e-wallet economy. Table 2 lists the two broad categories: Linked debit cards and credit cards constitute **external** options where e-wallets function as a conduit to banks and pass payment information to external ledgers; E-wallet balance, e-wallet savings, and BNPL are **internal** options where e-wallets function as an independent ecosystem and payment information stays within the wallets.

| | | (1) | (2) | (3) | (4) |
|--------------------|---------------|--------------|-------|--------------|--------|
| Types | Options | On-site | Total | Online | Total |
| | | transactions | share | transactions | share |
| External | Linked debit | 27.2% | | 16.0% | |
| (bank-linked) | card | 21.2% | 36.0% | 10.0% | 17.9% |
| options | Linked credit | 9.90/ | 30.0% | 1.00/ | 17.970 |
| | card | 8.8% | | 1.9% | |
| | E-wallet | 8.00/ | | 12.00/ | |
| Internal (e-wallet | balance | 8.0% | | 12.0% | |
| based) | E-wallet | 10.00/ | 62.7% | 15 40/ | 81.6% |
| options | savings | 10.9% | | 15.4% | |
| | BNPL | 43.8% | | 54.2% |] |

Table 2. Payment options within the e-wallet

Among external payment options, linked debit cards are the most frequently used, which is consistent with the low adoption of credit cards in China. Among internal payment options, the most popular is BNPL, which can be attributed to the opportunity costs associated with different internal payment options (e.g., Santomero and Seater, 1996; Klee, 2008): Specifically, because e-wallet savings offer positive interest rates, users prefer to exploit the free liquidity provided by BNPL for payments while leaving funds in e-wallet savings untouched to earn interest. In contrast, users seldom use e-wallet balances, as they provide neither interest nor additional liquidity. Note that Berg et al. (2023) find that BNPL dominates the payment choices (51%) for a large furniture merchant in Germany, which is consistent with our finding.

As shown in Table 2 Columns (2) and (4), the proportion of e-wallet internal payments is significantly larger than that of external payments for both on-site and

online transactions. Bank cards constitute only 36.0% and 17.9% of all on-site and online transactions, respectively. These results indicate that internal payment options dominate the e-wallet ecosystem, with e-wallet transactions largely bypassing banks. The relative distribution of internal versus external transactions holds important implications for the impact on traditional banking, as payment information is valuable and essential for other financial services such as credit scoring and origination (Agarwal et al., 2020a; Berg et al., 2022; Ghosh et al., 2022; Parlour et al., 2022). The rise of e-wallets and their dominant internal payment flows has led to banks being cut off from transaction information, motivating central banks worldwide, especially the People's Bank of China, to explore the introduction of their own digital currencies (known as central bank digital currencies, or CBDCs) to compete for payment flows (Auer et al., 2020; Boar and Wehrli, 2021; Brown et al., 2022; Whited, Wu, and Xiao, 2022). Our findings provide empirical support for these off-cited motivations for central bank electronic payment systems (EPSs) and CBDCs.

3.2 Payment Choice Model Specification

In examining the interaction between BNPL and other payment options and the financial inclusion effect of BNPL (as well as the subsequent analysis regarding the use of BNPL after incurring interest payments), we need to model the determinants of payment choices. For on-site transactions, we specify the following logit model:

$$Payment_option_dummy_{jh} = \alpha_h + \beta_1 D_Consumer_BNPL_j + \beta_2 D_Merchant_BNPL_j + \gamma X + \varepsilon_j$$
(1)

j denotes transaction, and *h* denotes five alternative payment options:

$$h \in \{e - wallet balance, e - wallet savings, BNPL, linked debit cards, linked credit cards\}.$$

Here, the dependent variable is *Payment_option_dummy*, which takes value of 1 when transaction *j* is completed with payment option *h*, and 0 otherwise. And α_h is a payment-option-specific constant term. *D_Consumer_BNPL* indicates whether the consumer in transaction *j* has access to BNPL, and *D_Merchant_BNPL* indicates whether the merchant in transaction *j* accepts BNPL. X represents a vector of control

variables: consumer access to credit cards ($D_Consumer_CC$); merchant acceptance of credit cards ($D_Merchant_CC$); whether the used payment option h is the default option of the consumer ($D_preferred_option$); how long the consumer has had access to payment option j (*Option_adoption_length*); the consumer's age, gender, the logarithm of e-wallet cash flows in the last year ($Ln_cashflow$), and the amount of wealth in the e-wallet (Ln_wealth); the logarithm of merchant's sales ($Ln_monthly_sales$); the logarithm of transaction value ($Ln_Transaction_value$). Industry fixed effects and daily time fixed effects are also included in the regression.

For online transactions, we adopt a similar econometric specification but omit the variables *D_Merchant_BNPL* and *D_Merchant_CC* because over 98% of online merchants accept both BNPL and credit cards:

Payment_option_dummy_{ih} =
$$\alpha_h + \beta_1 D_C consumer_B NPL_i + \gamma X + \varepsilon_i$$
 (2)

3.3 Impact of BNPL Access on Payment Choices

We first investigate how consumer access to and merchant acceptance of BNPL influence payment choices. Table 3 presents the results for on-site transactions. Note that consumer access to BNPL is significantly and negatively associated with all other payment option usage here, indicating that the expansion of e-wallet credit crowds out the usage of other payment options. Merchant acceptance of BNPL is positively associated with the usage of BNPL, credit cards, and e-wallet savings, but only the coefficient on BNPL usage is economically meaningful.

| | Payment options | | | | |
|------------------------------|-----------------|--------------|-----------|-----------|-------------|
| | E-wallet | Credit cards | E-wallet | E-wallet | Debit cards |
| | credit (BNPL) | | balance | savings | |
| | (1) | (2) | (3) | (4) | (5) |
| D_Consumer_BNPL ^d | 0.675*** | -0.021*** | -0.121*** | -0.157*** | -0.143*** |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| D_Consumer_CC ^d | 0.042*** | 0.063*** | -0.040*** | -0.040*** | 0.014*** |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| D_Merchant_BNPL ^d | 0.048*** | 0.001** | -0.002** | 0.007*** | -0.010*** |

Table 3. Payment-option choices for on-site transactions

| $ \begin{array}{llllllllllllllllllllllllllllllllllll$ | | (0.000) | (0.039) | (0.025) | (0.000) | (0.000) |
|---|---------------------------------|-----------|-----------|-----------|-----------|-----------|
| D_preferred_option ^d 0.339*** 0.021*** 0.112*** 0.227*** 0.268*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Option_adoption_length -0.000*** 0.000 -0.000*** -0.000*** 0.000) Age 0.002*** -0.001*** -0.001*** -0.005*** 0.000*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Female ^d -0.011*** 0.003*** -0.003*** -0.007*** 0.005*** (0.000) (0.000) (0.000) (0.000) (0.000) 0.005*** (0.000) (0.000) (0.001) (0.000) (0.000) Ln_transaction_value -0.033** 0.001*** 0.003*** 0.003*** 0.021*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) 0.004*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) 0.004*** Ln_transaction_value -0.015*** -0 | D_Merchant_CC ^d | -0.045*** | 0.049*** | -0.015*** | -0.011*** | 0.045*** |
| $ \begin{array}{cccccccccccccccccccccccccccccccccccc$ | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| $ \begin{array}{llllllllllllllllllllllllllllllllllll$ | D_preferred_option ^d | 0.339*** | 0.021*** | 0.112*** | 0.227*** | 0.268*** |
| Age (0.000) (0.390) (0.000) (0.000) (0.000) Age 0.002*** -0.001*** -0.001*** -0.005*** 0.000*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Female ^d -0.011*** 0.003*** -0.003*** -0.007*** 0.005*** (0.000) (0.000) (0.001) (0.000) (0.000) (0.000) (0.000) Ln_transaction_value -0.03*** 0.001*** 0.003*** 0.003*** 0.021*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_monthly_sales -0.015*** -0.001*** -0.005*** -0.006*** 0.004*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_cashflow -0.027*** -0.004*** 0.021*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_weath 0.010*** 0.000*** -0.024*** 0. | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Age0.002***-0.001***-0.001***-0.005***0.000***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Female ^d -0.011***0.003***-0.003***-0.007***0.005***(0.000)(0.000)(0.001)(0.000)(0.000)(0.000)Ln_transaction_value-0.033***0.001***0.003***0.003***0.021***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_monthly_sales-0.015***-0.001***-0.005***-0.006***0.004***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_cashflow-0.027***-0.004***0.020***0.012***-0.013***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_wealth0.010***0.000***-0.024***0.021***-0.008***(0.000)(0.004)(0.000)(0.000)(0.000)(0.000)Industry FEYESYESYESYESYESYESTime FEYESYESYESYESYESYESYES | Option_adoption_length | -0.000*** | 0.000 | -0.000*** | -0.000*** | 0.000*** |
| G (0.000) (0.000) (0.000) (0.000) (0.000) Female ^d -0.011*** 0.003*** -0.003*** -0.007*** 0.005*** (0.000) (0.000) (0.001) (0.000) (0.000) (0.000) Ln_transaction_value -0.033*** 0.001*** 0.003*** 0.003*** 0.021*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_monthly_sales -0.015*** -0.001*** -0.005*** -0.006*** 0.004*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_cashflow -0.027*** -0.004*** 0.020*** 0.012*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth VES YES YES | | (0.000) | (0.390) | (0.000) | (0.000) | (0.000) |
| Femaled-0.011***0.003***-0.003***-0.007***0.005***Ln_transaction_value-0.033***0.001***0.003***0.003***0.021***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_monthly_sales-0.015***-0.001***-0.005***-0.006***0.004***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_cashflow-0.027***-0.004***0.020***0.012***-0.013***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_wealth0.010***0.000***-0.024***0.021***-0.008***(0.000)(0.004)(0.000)(0.000)(0.000)(0.000)Industry FEYESYESYESYESYESYESTime FEYESYESYESYESYESYES | Age | 0.002*** | -0.001*** | -0.001*** | -0.005*** | 0.000*** |
| Ln_transaction_value(0.000)(0.000)(0.001)(0.000)(0.000)Ln_transaction_value-0.033***0.001***0.003***0.003***0.021***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_monthly_sales-0.015***-0.001***-0.005***-0.006***0.004***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_cashflow-0.027***-0.004***0.020***0.012***-0.013***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_wealth0.010***0.000***-0.024***0.021***-0.008***(0.000)(0.004)(0.000)(0.000)(0.000)(0.000)Industry FEYESYESYESYESYESYESTime FEYESYESYESYESYESYES | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_transaction_value -0.033*** 0.001*** 0.003*** 0.003*** 0.021*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_monthly_sales -0.015*** -0.001*** -0.005*** -0.006*** 0.004*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_cashflow -0.027*** -0.004*** 0.020*** 0.012*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.004) (0.000) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES YES Time FE YES YES YES YES YES YES | Female ^d | -0.011*** | 0.003*** | -0.003*** | -0.007*** | 0.005*** |
| Ln_monthly_sales(0.000)(0.000)(0.000)(0.000)(0.000)Ln_monthly_sales-0.015***-0.001***-0.005***-0.006***0.004***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_cashflow-0.027***-0.004***0.020***0.012***-0.013***(0.000)(0.000)(0.000)(0.000)(0.000)(0.000)Ln_wealth0.010***0.000***-0.024***0.021***-0.008***(0.000)(0.004)(0.000)(0.000)(0.000)(0.000)Industry FEYESYESYESYESYESYESTime FEYESYESYESYESYESYES | | (0.000) | (0.000) | (0.001) | (0.000) | (0.000) |
| Ln_monthly_sales -0.015*** -0.001*** -0.005*** -0.006*** 0.004*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_cashflow -0.027*** -0.004*** 0.020*** 0.012*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.004) (0.000) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES YES Time FE YES YES YES YES YES YES | Ln_transaction_value | -0.033*** | 0.001*** | 0.003*** | 0.003*** | 0.021*** |
| (0.000) (0.000) (0.000) (0.000) (0.000) Ln_cashflow -0.027*** -0.004*** 0.020*** 0.012*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES YES Time FE YES YES YES YES YES YES | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_cashflow -0.027*** -0.004*** 0.020*** 0.012*** -0.013*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.000) (0.000) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES YES Time FE YES YES YES YES YES YES | Ln_monthly_sales | -0.015*** | -0.001*** | -0.005*** | -0.006*** | 0.004*** |
| Image: Normal system (0.000) <td></td> <td>(0.000)</td> <td>(0.000)</td> <td>(0.000)</td> <td>(0.000)</td> <td>(0.000)</td> | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_wealth 0.010*** 0.000*** -0.024*** 0.021*** -0.008*** (0.000) (0.004) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES Time FE YES YES YES YES YES | Ln_cashflow | -0.027*** | -0.004*** | 0.020*** | 0.012*** | -0.013*** |
| (0.000) (0.004) (0.000) (0.000) (0.000) Industry FE YES YES YES YES YES Time FE YES YES YES YES YES YES | | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Industry FEYESYESYESYESTime FEYESYESYESYESYES | Ln_wealth | 0.010*** | 0.000*** | -0.024*** | 0.021*** | -0.008*** |
| Time FE YES YES YES YES | | (0.000) | (0.004) | (0.000) | (0.000) | (0.000) |
| | Industry FE | YES | YES | YES | YES | YES |
| Observations 501 801 501 801 501 801 501 801 501 801 | Time FE | YES | YES | YES | YES | YES |
| Julions Julion Julion Julion Julion Julion | Observations | 501,801 | 501,801 | 501,801 | 501,801 | 501,801 |
| Pseudo-R ² 0.25 0.41 0.20 0.15 0.14 | Pseudo-R ² | 0.25 | 0.41 | 0.20 | 0.15 | 0.14 |

Notes: 1. Subscript *d* indicates dummy variables. 2. *** p < 0.01; ** p < 0.05; * p < 0.1. 3.We report the marginal effects for the logit model.

As for control variables, as expected, consumers' payment-option preference is positively correlated with the payment-option usage, while option adoption length has little explanatory power. Older consumers are less likely to use e-wallet savings, perhaps reflecting that financial experience accumulated over time helps users to recognize the higher opportunity cost of e-wallet savings compared to the other options. Female consumers are less likely to use internal payment options and are more likely to use bank-linked external payment options. Debit cards are used more frequently in transactions with merchants with higher average monthly sales, which is in accordance with economic intuition since debit cards are still the more popular payment instrument accepted by large sellers. The cash flow and wealth variables in the e-wallet show mixed results. We find that the usage of BNPL is more likely for low-value transactions, while all other payment options are associated with higher-value transactions. Though we have only indirect measures of cash usage, the significant differences in coefficient estimation on transaction value for different payment options suggest how BNPL has become the new cash for small on-site transactions. Also note that the results are unlikely driven by the credit limits on BNPL because most e-wallet transactions did not exceed the BNPL credit limits (in the range of RMB ¥500-50,000). The annualized interest rate on delinquent payments for the BNPL credit is 13%-16%, which is slightly higher than the rates charged by most credit cards in China, ruling out the possibility that consumers embrace BNPL due to lower interest rates.

Next, we turn to online transactions. As shown in Table 4, consumer access to BNPL similarly reduces the possibility of all other payment-option usage, indicating that e-wallet credit access as a payment option crowds out the other options.

Regarding transaction value, the results are more mixed for online transactions than for on-site transactions. The coefficients on transaction value are positive for bank cards, including both credit cards and debit cards, and are negative for e-wallet balance and e-wallet savings. While credit card usage is still positively associated with transaction value, based on the negative but insignificant coefficient on transaction value in Column (1), we find no evidence to suggest that e-wallet credit is used for big-ticket purchases.

| | Payment opti | Payment options | | | | |
|------------------------------|--------------|-----------------|-----------|-----------|-----------|--|
| | E-wallet | Credit cards | E-wallet | E-wallet | Debit | |
| | credit | | balance | savings | cards | |
| | (BNPL) | | | | | |
| | (1) | (2) | (3) | (4) | (5) | |
| D_Consumer_BNPL ^d | 0.603*** | -0.102*** | -0.079*** | -0.099*** | -0.149*** | |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) | |
| D_Consumer_CC ^d | -0.030*** | 0.311*** | -0.063*** | -0.062*** | -0.130*** | |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) | |

Table 4. Payment-option choices for online transactions

| D_preferred_option ^d | 0.324*** | 0.130*** | 0.055*** | 0.161*** | 0.321*** |
|---------------------------------|-----------|-----------|-----------|-----------|-----------|
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Option_adoption_length | 0.000*** | -0.000*** | -0.000*** | -0.000*** | 0.000*** |
| | (0.001) | (0.000) | (0.000) | (0.005) | (0.004) |
| Age | 0.001*** | -0.000*** | -0.000 | -0.002*** | -0.001*** |
| | (0.000) | (0.000) | (0.143) | (0.000) | (0.000) |
| Female ^d | -0.010*** | 0.003*** | -0.009*** | -0.010*** | 0.014*** |
| | (0.000) | (0.000) | (0.001) | (0.000) | (0.000) |
| Ln_transaction_value | -0.001 | 0.002*** | -0.009*** | -0.010*** | 0.024*** |
| | (0.119) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_monthly_sales | -0.004*** | -0.005*** | -0.006*** | -0.005*** | -0.001*** |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_cashflow | -0.005*** | -0.007*** | 0.016*** | 0.010*** | -0.020*** |
| | (0.000) | (0.000) | (0.000) | (0.000) | (0.000) |
| Ln_wealth | 0.010*** | 0.000*** | -0.016*** | 0.019*** | -0.016*** |
| | (0.000) | (0.004) | (0.000) | (0.000) | (0.000) |
| Industry FE | YES | YES | YES | YES | YES |
| Time FE | YES | YES | YES | YES | YES |
| Observations | 501,801 | 501,801 | 501,801 | 501,801 | 501,801 |
| Pseudo-R ² | 0.25 | 0.41 | 0.20 | 0.15 | 0.14 |

Notes: 1. Subscript *d* indicates dummy variables. 2. *** p < 0.01; ** p < 0.05; * p < 0.1. 3.We report the marginal effects for the logit model.

In sum, we find significant evidence that BNPL access displaces other payment options.²⁶ We also find that BNPL usage is not competing with credit cards for large value purchases, even though it resembles a virtual credit card.

3.4 BNPL vs Credit Cards: Competition and Differentiation

BNPL crowds out other payment choices, but the competition between BNPL and credit cards is nuanced and warrants further investigation. In this section, we

²⁶ Though not included in the main text, we perform several robustness tests regarding the results in Tables 3 and 4. First, we apply the probit model and ordinary least squares method to estimate the regression model. Second, in Eqs. (1) and (2), we use a dummy variable indicating whether the used payment option is the most preferred (default) one ($D_preferred_option$) to proxy for consumer preference. We instead use the proportion of transactions using a payment option over the last three months as an alternative proxy. Third, we combine the online and on-site datasets to jointly estimate the regression model following Eq. (1). Our findings remain robust. Fourth, we re-run the regressions with city or province fixed effects. The logit model did not converge due to incidental parameter problems. We instead apply OLS to the model with city or province fixed effects, and our findings remain unchanged. For example, with city fixed effects, the coefficient estimates on $D_consumer_BNPL$ for on-site transactions are statistically significant at the 1% level, with the magnitudes of 0.279, -0.011, -0.192, -0.163, and -0.211 in the five columns respectively. The results are available upon request.

distinguish between *single access* users who have access only to BNPL and not credit cards, and *dual access* users with access to both BNPL and credit cards. While most e-wallet users only have single access to BNPL and do not involve competition between the two payment methods, an intriguing question remains for dual-access users: given BNPL access crowds out credit cards as payment options, especially on-site, is e-wallet credit a substitute for bank credit?

Previous studies compare the patterns of FinTech borrowers with similar borrowers from traditional financial institutions (e.g., Jagtiani and Lemieux, 2019; Di Maggio and Yao, 2021). The data limitations could result in potential bias due to borrowers' unobservable characteristics; in other words, borrowers who have access to FinTech credit may be inherently different from borrowers with access to credit from traditional financial institutions. The ideal empirical condition is to have detailed information on borrowers who have both FinTech credit and bank credit and compare their usage of these two types of credit, which is the case in the subsample of dual-access users.

Table 5. Consumer credit as payment instruments for on-site transactions:

| User type: | Single-access users | Dual-access users | | |
|----------------------------------|---------------------|-------------------|---------------|--|
| Dependent variable: whether a | BNPL=1 | BNPL=1 | Credit card=1 | |
| payment instrument is used in an | | | | |
| on-site transaction | (1) | (2) | (3) | |
| D_Merchant_BNPL ^d | 0.026*** | 0.037*** | -0.010*** | |
| D_merchant_Bive L | (0.000) | (0.000) | (0.000) | |
| D Marcont CCd | -0.051*** | -0.104*** | 0.197*** | |
| D_Mercent_CC ^d | (0.000) | (0.000) | (0.000) | |
| Commune and former and | 0.391*** | 0.358*** | 0.105*** | |
| Consumer_preference ^d | (0.000) | (0.000) | (0.000) | |
| Ontion edention longth | 0.000 | 0.000 | 0.000*** | |
| Option_adoption_length | (0.858) | (0.218) | (0.000) | |
| C | 0.003*** | 0.005*** | -0.002*** | |
| Consumer age | (0.000) | (0.000) | (0.000) | |
| Female ^d | -0.006*** | 0.002 | -0.001 | |
| | (0.000) | (0.359) | (0.579) | |
| In transaction value | -0.029*** | -0.048*** | 0.005*** | |
| Ln_transaction_value | (0.000) | (0.000) | (0.000) | |

single-access users vs. dual-access users

| Ln_monthly_sales | -0.005 *** | -0.007 * * * | -0.002*** |
|-----------------------|------------|--------------|-----------|
| Ln_monuny_sales | (0.000) | (0.000) | (0.000) |
| I | -0.029*** | -0.020*** | -0.008*** |
| Ln_cashflow | (0.000) | (0.000) | (0.000) |
| I n waalth | 0.010*** | 0.010*** | -0.001*** |
| Ln_wealth | (0.000) | (0.000) | (0.000) |
| Industry FE | YES | YES | YES |
| Day FE | YES | YES | YES |
| Observations | 304,137 | 144,133 | 144,133 |
| Pseudo-R ² | 0.13 | 0.13 | 0.38 |

Note: 1. The superscript *d* indicates a dummy variable. 2. *** p < 0.01; ** p < 0.05; * p < 0.1. 3.We report the marginal effects for the logit model.

In Table 5, we present the determinants of the usage of consumer credit (e-wallet credit through BNPL or bank credit cards) as the payment options in on-site transactions for single-access users and dual-access users, respectively. Dual-access users exhibit a clear pattern of using BNPL for low-value purchases and credit cards for big-ticket purchases (based on the contrasting coefficient estimation on transaction value in Columns (2) and (3)). For single-access users, transaction value is negatively associated with the usage of BNPL in on-site transactions. Considering this, BNPL appears not to be a direct substitute for bank credit cards but rather serves as cash for a separate segment of transactions.

We repeat the analyses for online transactions and report the results in Table 6. Similarly, for dual-access users, BNPL usage is associated with lower transaction values, while credit card usage is associated with high-value purchases. In comparison, single access users do not exhibit the same usage patterns. The results in Table 6 further corroborate the conjecture that BNPL and credit cards have different usage patterns and are not direct substitutes at this stage of development.

There may be multiple reasons for the observed differences in usage patterns between BNPL and credit cards. One important factor could be the greater acceptance of BNPL by merchants, particularly those who do not accept credit cards. However, this factor alone cannot fully explain the differences observed in online transactions, because most online merchants accept both payment methods. Two other potential factors are: (1) BNPL credit limits are typically lower than those of credit cards, which may necessitate the use of credit cards for larger purchases; (2) frequent usage of BNPL for small transactions at checkout may spill over to online transactions, resulting in greater use of BNPL for such transactions.

The competition between BNPL and credit cards is nuanced. The majority of BNPL users are single-access users who did not have prior access to credit cards. Transaction size also matters: our analysis indicates that dual-access users primarily utilize BNPL for small payments and prefer credit cards for larger purchases. While competition between BNPL and credit cards may exist for middle-value purchases, BNPL and credit card usage have diverged to cater to users in different purchases.²⁷ In Section 5, we also provide evidence that providing BNPL access to credit card holders leads to an increase in credit card consumption. In this sense, e-wallet credit can complement bank credit.

| User type: | Single-access users | Dual-a | ccess users | |
|----------------------------------|---------------------|-----------|---------------|--|
| Dependent variable: whether a | BNPL=1 | BNPL=1 | Credit card=1 | |
| payment instrument is used in | | | | |
| an online transaction | (1) | (2) | (3) | |
| Commun and | 0.355*** | 0.440*** | 0.407*** | |
| Consumer_preference ^d | (0.000) | (0.000) | (0.000) | |
| Option_adoption_length | 0.000*** | -0.000 | 0.000** | |
| | (0.000) | (0.962) | (0.036) | |
| Communication | 0.002*** | 0.001*** | 0.002*** | |
| Consumer age | (0.000) | (0.000) | (0.000) | |
| Female ^d | -0.011*** | -0.012*** | 0.021*** | |
| | (0.000) | (0.000) | (0.000) | |
| I a turneration relation | -0.000 | -0.005*** | 0.019*** | |
| Ln_transaction_value | (0.896) | (0.000) | (0.000) | |
| | | | | |

Table 6. Consumer credit as payment instruments for online transactions: single-access users

vs. dual-access users

 $^{^{27}}$ To rule out the possibility that the crowding-out effect of BNPL on credit cards is driven by dual-access users who have already exhausted their credit card limits, we repeated the analyses in Column (2) of Tables 3 and 4 using a subsample of users whose e-wallet savings amounts are above the median, and thus have de facto access to credit cards. The results show that the crowding-out effect remains significant at the 1% level, with coefficients on *D* Consumer BNPL of -1.607 and -2.094 for on-site and online transactions, respectively.

| In monthly sales | 0.005*** | -0.001** | 0.001 |
|-----------------------|-----------|----------|-----------|
| Ln_monthly_sales | (0.000) | (0.011) | (0.223) |
| In aashflaw | -0.007*** | 0.000 | -0.014*** |
| Ln_cashflow | (0.000) | (0.791) | (0.000) |
| I n wealth | 0.011*** | 0.010*** | -0.005*** |
| Ln_wealth | (0.000) | (0.000) | (0.000) |
| Industry FE | YES | YES | YES |
| Day FE | YES | YES | YES |
| Observations | 281,300 | 151,836 | 151,836 |
| Pseudo-R ² | 0.09 | 0.11 | 0.08 |

Notes: 1. The superscript *d* indicates a dummy variable. 2. *** p < 0.01; ** p < 0.05; * p < 0.1. 3.We report the marginal effects for the logit model.

3.5 Competition Beyond the E-Wallet: BNPL as the New Cash?

The long-standing discussion on the "curse of cash" questions whether technological advances can effectively reduce cash usage. Although our analysis does not directly address physical cash since as it is an option outside e-wallets, our findings suggest that BNPL has the potential to become the new cash for small-valued daily payments.

Our study incorporates the most convenient forms of digital payments via e-wallets and allows us to discuss what has become the new cash in the digital age. While earlier studies document that cash is heavily used when available (Alvares and Argente, 2022) and electronic payment has limited effects on cash demand (Brown et al., 2022), these studies do not cover more advanced electronic payment systems or economies that have transitioned from a cash-heavy society to a cashless society. As recently as 2006, cash use in China still accounted for 13% of GDP, much higher than that in the US (6.4%) and the UK (3.5%). However, the share of cash used at point of sale (POS) in China has dropped from 74.7% to 25.4% between 2012 and 2020, making China one of the most cashless countries.²⁸

We show that BNPL performs the function of cash in settling small-value transactions. To this end, Ho et al. (2022) find that the share of mobile payments has increased over time at the expense of cash payments, attributing this increase to price

²⁸ Accessed from https://english.ckgsb.edu.cn/knowledges/will-china-first-cashless-society/ and https://www.statista.com/statistics/1306790/cash-use-in-china/.

incentives. We propose additional driving forces with a more recent perspective: expanded consumer access to, and merchant acceptance of, e-wallet credit.

Another important functionality of cash is enabling the avoidance of tax payments. Over the past decade, e-wallet usage has also largely served this purpose, as merchants could partially avoid taxes because e-wallets are not "business bank accounts" with direct tax information. That said, there have been lawsuit cases against merchants that reported sales too low to be realistic, and regulators are tightening the to scrutiny to require e-wallet providers report transactions with a single-day-cumulative value larger than certain thresholds for tax purposes. Unlike cash, BNPL and other e-wallet payment options are not as private, leaves digital footprints, and cannot be used for various illicit activities. For these reasons, cash will likely continue to be used in situations where privacy and anonymity are required. Whether these situations are harmful to society is a separate question.

Finally, since an e-wallet super app ecosystem fundamentally differs from a pure-play payment network based on bank cards, different payment options may lead to different merchant fees, retail prices, and consumer usage segments. As a result, in contrast to the findings documented in Wang (2023), which uses U.S. bank card data and focused on more homogenous payment networks, merchant fees do not increase or get passed onto retail prices, and consumers do not appear to overuse the newly available credit (as we demonstrate in Section 5). In an economy transitioning from a cash-heavy to cashless system, the introduction of e-wallets and BNPL appears to facilitate financial inclusion rather than reduce welfare, as we discuss next.

4. Financial Inclusion via E-Wallet Credit Expansion

The competitive landscape of payment options documented above raises the question: has e-wallet credit through BNPL displaced incumbents such as traditional banks, or has it filled the gap in an underserved credit market? This section addresses this question. Since e-wallet providers can leverage payment information to evaluate users' creditworthiness, we conjecture that they are expanding credit access to those underserved by traditional banks. In this section, we provide some suggestive evidence for e-wallet credit expansion at the extensive margin.

One core question studied in the FinTech-lending literature concerns whether FinTech credit is offered to inframarginal borrowers or borrowers with no prior credit access or score. Due to data limitations, previous studies cannot directly observe bank-credit access and FinTech-credit access for the same individual at the same time. They therefore had to indirectly infer the composition of FinTech borrowers through structural models (Tang, 2019) or characteristic matching (Jagtiani and Lemieux, 2019; Di Maggio and Yao, 2021). We can overcome this empirical challenge because we can directly observe whether FinTech borrowers, i.e., e-wallet users of BNPL, have access to bank credit.

One may argue that e-wallet users may choose not to link their credit cards to e-wallet accounts, which would result in selection bias. We acknowledge this possibility but assuage the concern in several ways. First, e-wallet users are required to link at least one bank card when opening an account. Second, given the prevalence of e-wallets and QR payments in China, users are motivated to link their credit cards to improve their e-wallet credit scores, to enjoy credit card–related services provided by e-wallets, and to enable mobile front-end usage of their credit cards.²⁹ It is therefore highly improbable that users choose not to link credit cards to their e-wallets. Third, in our sample, the fraction of e-wallet users with linked credit cards and the number of credit cards owned among Chinese adults overall—around 30% and 0.56, respectively—which means e-wallet users are representative of Chinese credit card ownership.³⁰ Fourth, having an unlinked credit card would not bias against most of

²⁹ For example, one can pay credit card debt through e-wallet apps. And usually, e-wallet apps outperform bank apps in loading speed and user-friendliness.

 $^{^{30}}$ More than two-thirds of adults in China do not have a credit history and credit score. See http://www.gov.cn/xinwen/2015–10/27/content_2954607.htm. Note that e-wallet users tend to link most credit cards they own. The average number of credit cards linked per person in our sample is around 0.55, close to the number held per person as reported by the People's Bank of China, which is 0.56. See

our findings. Finally, we compare borrower quality following Tang (2019), if the average borrower quality of those without credit cards is lower compared with borrowers with linked credit cards, we can draw a more reliable conclusion that credit has expanded at the extensive margin.

It is worth noting that the richness of the data enables us to add a new perspective on the expansion of FinTech credit: the merchant's side. Previous studies on FinTech credit did not need to consider the merchant side. This is because they mainly focused on marketplace lending, which is not directly influenced by merchant acceptance. However, the new type of FinTech credit considered here provides credit lines and installment options for purchases made by consumers; hence, merchant acceptance of the payment option is an important factor in evaluating the extent of credit expansion. Consumers' access to credit would be negligible if few merchants accepted BNPL.

We first analyze credit expansion at the extensive margin from the consumer side, and then we analyze the expansion from the merchant side. We also combine the consumer side and the merchant side to determine whether a transaction could be completed with credit. Based on that, we provide an estimation of efficient credit expansion in this two-sided payment network. Importantly, the indirect network effect implies that the benefits users derive from a payment option depend on supply-side adoption of the technology to accept this payment option, which in turn relies on the number of consumers adopting this payment option and the amount of merchant fees the e-wallet provider charges.³¹ Moreover, we conduct a user-quality comparison between e-wallet credit and linked credit cards to demonstrate how e-wallets promote financial inclusion at the extensive margin. Last, we compare the utilization of BNPL and credit cards in different regions of China with varying economic conditions, or along other dimensions of user heterogeneity (e.g., gender), which further corroborate the financial inclusion effect of BNPL.

http://www.pbc.gov.cn/goutongjiaoliu/113456/113469/4213347/2021032414491874847.pdf.

³¹ Higgins (2019) attempts to study the magnitude of these externalities and resulting spillovers of financial technology adoption within and across the two sides of the market using data from a large-scale conditional cash transfer program, Prospera, in Mexico.

4.1 Credit Expansion for Consumers

Table 7 shows the credit expansion from the consumer side. First, as shown in Panel A, among on-site consumers, 70.43% do not have a credit card linked to their e-wallets. For online consumers, the percentage is 68.31%. However, 84.14% of those with no credit card for on-site consumers and 78.38% for online consumers utilize BNPL as a form of e-wallet credit. This contrasts with some findings about the U.S. FinTech market, in which FinTech lenders primarily cater to borrowers who already have access to credit via traditional banks (Tang, 2019; Di Maggio and Yao, 2021).

Second, as shown in Panel B, among the 29.57% of on-site consumers with linked credit cards, 94.93% also enjoy e-wallet credit access. For the 31.69% of online consumers with linked credit cards, 90.91% have e-wallet credit access. This indicates that e-wallets entice bank customers to adopt the new credit line with BNPL, in line with the finding of Tang (2019) and Di Maggio and Yao (2021) that FinTech lenders usually increase their market share by extending credit to bank borrowers.

| Consumers | On-site | Online |
|--|---------|--------|
| Panel A: Without credit cards | 70.43% | 68.31% |
| % of the above with e-wallet credit (BNPL) | 84.14% | 78.38% |
| Panel B: With credit cards | 29.57% | 31.69% |
| % of the above with e-wallet credit (BNPL) | 94.93% | 90.91% |
| Panel C: With e-wallet credit (BNPL) | 87.32% | 82.28% |
| % of the above with credit cards | 32.14% | 35.02% |

Table 7. Expansion of e-wallet credit from consumers' side

Third, as shown in Panel C, 87.32% of on-site e-wallet users have e-wallet credit, while 82.28% of online e-wallet users have e-wallet credit. However, only 32.14% of on-site users and 35.02% of online users have linked credit cards at the same time. This finding indicates that the majority of BNPL users do not have linked credit cards. Compared to the inframarginal credit expansion documented in Tang (2019), the

credit expansion at the extensive margin—that is, expanding credit access to those with no access to bank credit—is more pronounced in our sample.

We attribute the differences between our findings and those of previous studies to the differences in types of FinTech credit providers and their information advantages compared to traditional banks. Tang (2019) and Di Maggio and Yao (2021) focus on FinTech credit provided by online lenders and P2P lenders such as Lending Club. In contrast, our study examines FinTech credit granted by e-wallet providers. While P2P lenders do not have significant information advantages compared to traditional banks, e-wallet providers enjoy rich information from consumers' digital footprints and online purchase histories; thus, e-wallet providers are able to provide credit to those underserved by banks. As such, BNPL providers can have a more pronounced credit-expansion effect than P2P lenders at the extensive margin. As the old FinTech lending model (P2P) has disappointed and the most notable new FinTech business model (BNPL) is developing rapidly (Berg, Fuster, and Puri, 2022), the facts we document help assess the implications of the recent structural change in the FinTech lending landscape.

4.2 Merchants' Acceptance of BNPL

Table 8 shows credit expansion from the merchant side. In Column 1 of Table 8, Panels A, B, and C report credit expansion for on-site merchants. As shown in Panel A, among on-site merchants, 81.79% do not accept credit card payments. However, 43.54% of them accept BNPL, possibly because of its convenience and large user base. Next, as illustrated in Panel B, among the 18.21% of merchants who accept credit cards, 83.42% of them choose to simultaneously embrace BNPL. Lastly, as demonstrated in Panel C, more than half (50.81%) of merchants accept BNPL payments, and only 29.90% of them also accept credit cards, and according to the e-wallet provider, they have never accepted credit cards previously, likely because for BNPL they need to neither supply the credit nor pay an exorbitant fee to the

intermediaries. In addition, they can enjoy the larger customer base of BNPL. Therefore, the extensive marginal effect, not the intensive marginal effect, dominates the expansion of e-wallet credit for on-site merchants.

| Merchants | On-site | Online |
|---|---------|--------|
| Panel A: Do not accept credit cards | 81.79% | 1.08% |
| % of the above that accept e-wallet credit (BNPL) | 43.54% | 80.56% |
| Panel B: Accept credit cards | 18.21% | 98.92% |
| % of the above that accept e-wallet credit (BNPL) | 83.42% | 99.37% |
| Panel C: Accept e-wallet credit (BNPL) | 50.81% | 99.17% |
| % of the above that accept credit cards | 29.90% | 99.12% |

Table 8. The expansion of e-wallet credit from merchants' side

In Column 2 of Table 8, Panels A, B, and C demonstrate credit expansion for online merchants. Unlike on-site merchants, nearly all online merchants accept credit card payments, with 98.92% as shown in Panel B. Although the acceptance ratio of BNPL is high at 99.17%, as shown in Panel C, BNPL's credit expansion at the extensive margin for online merchants is tiny. Since 98.30% of all online merchants accept both types of credit payments, the intensive marginal effect likely dominates the expansion of BNPL for online merchants.

One may be concerned that some merchants accepting credit cards would stop doing so once they adopt BNPL. According to the e-wallet provider, this occurrence is extremely rare, because many consumers, especially the ones with large transactions, still use credit cards. ³² Merchants who find credit cards unattractive would also opt for cash for on-site transactions to start with, rather than switching to BNPL. For online merchants, such a switch is virtually non-existent because 98.92% of them accept credit cards.

4.3 Efficient Credit Expansion: Evidence from Transactions

³² To be specific, less than 1% of the merchants who previously accepted credit cards stopped doing so after their adoption of BNPL in the entire history of the e-wallet provider. Therefore, the observed merchants who adopt BNPL but not credit cards represent an extensive margin for credit expansion through BNPL.

This subsection discusses efficient credit expansion by combining the merchant side and the consumer side. A transaction can be completed using BNPL only if the consumer has access to it and the merchant accepts it as a payment method. Thus, credit expansion is efficient only when both the consumer's access to a payment method (using credit) and the merchant's acceptance of that payment method are aligned.

To measure efficient credit expansion, we first determine whether a transaction can be completed with credit cards or BNPL based on both merchant acceptance and consumer access. Next, we classify the transactions into different categories: those that can be completed with credit cards only, those that can be completed with either BNPL or credit cards, and those that can be completed with BNPL only. When a transaction can be completed with BNPL only, we deem it an efficient credit expansion.

Table 9 shows that efficient credit expansion remains significant along various dimensions. First, as illustrated in Panel A, 91.59% of on-site and 64.44% of online transactions cannot be completed with credit cards. However, 44.37% and 90.21% of them can be completed with BNPL. Hence, the extensive margin of efficient credit-access expansion is economically meaningful. Next, as demonstrated in Panel B, among the 8.41% of on-site transactions that can be completed with credit cards, 75.62% can also be completed with BNPL. For the 31.56% of online transactions that can be completed with credit cards, 90.46% can also be completed with BNPL. Lastly, as revealed in Panel C, among the 47.00% of on-site transactions completed with BNPL, only 13.53% can be completed with BNPL, only 34.95% can be completed with credit cards. The findings suggest that the extensive marginal effect, not the intensive marginal effect, dominates efficient credit expansion at the transaction level.

Table 9. Efficient credit expansion: combining consumers' and merchants' sides

| Transactions | On-site | Online |
|--|---------|--------|
| Panel A: Cannot use credit cards | 91.59% | 64.44% |
| % of the above that can use e-wallet credit (BNPL) | 44.37% | 90.21% |
| Panel B: Can use credit cards | 8.41% | 31.56% |
| % of the above that can use e-wallet credit (BNPL) | 75.62% | 90.46% |
| Panel C: Can use e-wallet credit (BNPL) | 47.00% | 81.68% |
| % of the above that can use credit cards | 13.53% | 34.95% |

Furthermore, we follow Tang (2019) and make a simple comparison of user quality. We have two reasons to believe that e-wallets are expanding credit to users of lower borrower quality compared with bank credit card holders. One reason, mentioned in previous studies like Tang (2019), is that e-wallet providers can adopt laxer lending standards due to several factors such as lower operation costs. Another, somewhat subtler, reason that cannot be ignored in the two-sided payment market is that the merchants who accept credit card payments are usually large malls and well-known e-commerce brands that sell big-ticket products; as a result, young consumers with low income have little chance to make purchases with credit cards. However, e-wallet providers have greatly promoted credit acceptance among merchants, including grocery stores and startup brands. As merchants increasingly accept BNPL, younger consumers with lower income can conveniently use BNPL in their daily shopping activities. Thus, the two-sided nature of the market and the high rate of merchant acceptance provide another explanation of why e-wallet credit is helping users of lower credit quality.

| | (1) Dual-access users | (2) Single-access users |
|--------------------------|-----------------------|-------------------------|
| Panel A: on-site | | |
| Age | 33.77 | 32.27 |
| Wealth in e-wallet (RMB) | 8306.88 | 5935.97 |
| Average City tier level | 2.91 | 3.22 |
| Panel B: online | | |
| Age | 33.12 | 31.23 |
| Wealth in e-wallet (RMB) | 9948.66 | 6447.88 |
| Average City tier level | 2.89 | 3.40 |

Table 10. User-quality comparison of efficient credit-access expansion

Notes: "Average city tier level" refers to the average value of the city tier level of users' locations. In China, cities are classified into different tiers based on factors such as population, GDP, and infrastructure development. For example, first-tier cities such as Beijing, Shanghai, Guangzhou, and Shenzhen are among the most developed cities

in China, while second-tier cities such as Chengdu are less developed than first-tier cities but more developed than third- or fourth-tier cities.

Table 10 compares the quality of dual-access and single-access users for on-site and online transactions. Panel A demonstrates that dual-access users are older, wealthier, and located in more developed cities than single-access users. This suggests that e-wallet credit is primarily used by consumers who have limited access to bank credit or young consumers with low income. We reach similar conclusions concerning online transactions from Panel B.

4.4 Financial Inclusion Through BNPL? Evidence from Regional Usage

This subsection further corroborates the credit expansion and financial inclusion role of BNPL. We determine the location of each transaction by identifying the consumer's place of residence and split the sample into well-developed and less-developed regions of China. We then compare the usage of BNPL and credit cards across these regions to examine the extent to which BNPL has facilitated financial inclusion.

Table 11 presents the comparison results. Panel A shows the results categorized by urban and rural areas, and Panel B examines the comparison between the southern and northern regions. Relative to northern regions, southern regions are more heavily populated areas with better socioeconomic development (Huang et al., 2020).

Table 11. Usage comparison of BNPL and credit cards in different regions

| | Fraction of on-site transactions using BNPL and credit cards | | Fraction of online transactions using BNPL and credit cards | | | | |
|----------------------|--|------------------|---|------------------|--|--|--|
| | BNPL (%) | Credit cards (%) | BNPL (%) | Credit cards (%) | | | |
| Panel A | Panel A | | | | | | |
| Urban area (1) | 54.30 | 2.36 | 43.65 | 10.97 | | | |
| Rural area (2) | 54.30 | 1.09 44.34 | | 5.24 | | | |
| (1)-(2) | 0.00 | 1.27*** | -0.69*** | 5.72*** | | | |
| P value | 0.99 | 0.00 0.00 | | 0.00 | | | |
| Panel B | · | · | | · | | | |
| Southern regions (7) | 52.17 | 2.04 | 43.68 | 8.91 | | | |
| Northern regions (8) | 58.24 | 1.69 | 43.91 | 8.68 | | | |
| (7)-(8) | -6.07*** | 0.35*** | -0.23 | 0.23*** | | | |
| P value | 0.00 | 0.00 | 0.10 | 0.00 | | | |

Notes: *** p < 0.01; ** p < 0.05; * p < 0.1

As expected, the proportion of transactions using credit cards in more developed regions (urban areas and southern regions) is significantly higher than that in less developed regions (rural areas and northern regions). In contrast, BNPL, as another consumer credit, does not exhibit a similar pattern. The proportion is even higher among on-site transactions in northern regions and among online transactions in rural areas. In this sense, the credit expansion of BNPL alleviates the inequality in access of bank consumer credit between well-developed and less-developed areas in China, consequently promoting financial inclusion.

We further explore the determinants of consumer credit usage (credit cards and BNPL) as a whole, with a special focus on the heterogeneous impacts of consumer access to BNPL in well-developed and less-developed areas. To achieve this, we introduce two dummy variables (*Rural* and *North*) indicating whether a transaction occurs in rural areas or northern regions, and their interactions with consumer access

| | The dependent variable: =1 if a transaction is conducted by consumer credit (BNPL or credit | | | |
|------------------------------|---|----------|---------------------|----------|
| | cards) | | | |
| | On-site transactions | | Online transactions | |
| | (1) (2) | | (3) | (4) |
| D_Consumer_BNPL ^d | 0.443*** | 0.437*** | 0.205*** | 0.220*** |

Table 12. The determinants of consumer credit usage in different regions

| | (0.000) | (0.000) | (0.000) | (0.000) |
|--|-----------|---------------------------|--------------|-----------|
| D_Consumer_CC ^d | 0.086*** | 0.083*** | 0.233*** | 0.220*** |
| | (0.000) | (0.000) | (0.000) | (0.000) |
| D_Merchant_BNPL ^d | 0.018*** | 0.019*** | (0.000) | (|
| | (0.000) | (0.000) | | |
| D_Merchant_CC ^d | 0.030*** | 0.020*** | | |
| | (0.000) | (0.000) | | |
| D_preferred_option ^d | 0.340*** | 0.342*** | 0.283*** | 0.284*** |
| | (0.000) | (0.000) | (0.000) | (0.000) |
| Rural ^d | -0.119*** | | -0.139*** | |
| | (0.000) | | (0.000) | |
| D_Consumer_BNPL ^d * | | | | |
| Rural ^d | 0.113*** | | 0.142*** | |
| | (0.000) | | (0.000) | |
| $D_Consumer_CC^d * Rural^d$ | -0.003 | | -0.064*** | |
| | (0.248) | | (0.000) | |
| D_Merchant_BNPL ^d * | | | | |
| Rural ^d | 0.037*** | | | |
| | (0.000) | | | |
| D_Merchant_CC ^d * Rural ^d | -0.077*** | | | |
| | (0.000) | | | |
| North ^d | | -0.097*** | | -0.076*** |
| | | (0.000) | | (0.000) |
| D_Consumer_BNPL ^d * | | | | |
| North ^d | | 0.116*** | | 0.075*** |
| | | (0.000) | | (0.000) |
| D_Consumer_CC ^d * North ^d | | -0.000 | | -0.006** |
| | | (0.870) | | (0.024) |
| D_Merchant_BNPL ^d * North ^d | | 0.025*** | | |
| North | | 0.035*** | | |
| D_Merchant_CC ^d * North ^d | | (0.000) -0.059*** | | |
| D_Merchant_CC • Norm | | (0.000) | | |
| Control variables | YES | YES | YES | YES |
| Industry FE | YES | YES | YES | YES |
| Day FE | YES | YES | YES | YES |
| Pseudo R ² | 0.229 | 0.230 | 0.219 | 0.218 |
| Observations | 501,598 | 501,801 | 512,505 | 512,803 |
| | | 0. database 0. 0.01 - 1.1 | 0.05 * 0.1.2 | 512,005 |

Notes: 1. Subscript *d* indicates dummy variables. 2. *** p < 0.01; ** p < 0.05; * p < 0.1. 3.We report the marginal effects for the logit model.

to BNPL (*D_Consumer_BNPL*). Additionally, for completeness, we incorporate the interaction terms between *Rural* (or *North*) and consumer access to credit cards (*D_Consumer_CC*), merchant access to BNPL (*D_Merchant_BNPL*), and merchant

access to credit cards ($D_Merchant_CC$). Table 12 reports the regression results, where the dependent variable is an indicator representing whether consumer credit (credit card or BNPL) is used in a transaction. First, the coefficients on $D_Consumer_BNPL$ are positive and statistically significant at the 1% level, indicating that consumer access to BNPL improves the use of consumer credit in transactions in more-developed regions (urban areas and southern regions). Second, the coefficients on the interaction terms on $D_Consumer_BNPL$ and Rural (or North) are also positive and statistically significant, indicating that consumer access to BNPL has a larger impact on consumer credit usage in less-developed regions (rural areas and northern regions).

In sum, considering the user quality and the two-sided nature of the market, we conclude that the expansion of e-wallet credit is economically significant at the extensive margin and BNPL is mostly provided to those with no access to bank credit: about 80% of consumers, 44% of on-site merchants, 44% of on-site transactions, and 90% of online transactions that cannot be completed by credit cards can now be completed with BNPL. The credit expansion through BNPL seems to have promoted financial inclusion, especially in less-developed regions and among women.³³

5. The Impact of BNPL on Consumer Spending and Indebtedness

BNPL has displaced other payment choices, exhibited payment usage patterns different from credit cards, and expanded credit at the extensive margin. Next, we investigate its impacts on consumer spending, default, and indebtedness.

The canonical life-cycle/permanent-income hypothesis states that an increase in predictable income or credit access should have no real effects on consumption. However, this conclusion does not hold when consumers are financially or liquidity

³³ Though unreported here, we also examine the heterogeneous impact of BNPL access on consumer credit usage in the dimension of consumers' gender. Specifically, the coefficients for the female dummy variable and its interaction with *D_Consumer_BNPL* are significant at the 1% level, with magnitudes of -0.815 and 0.523 for on-site transactions and -1.165 and 1.008 for online transactions. These results indicate that women face disadvantages in credit access and usage, but BNPL access helps them more.

constrained. Yet, indirect measures for financial or liquidity constraints usually conflate credit demand and supply. Following Gross and Souleles (2002) and Aydın (2022) that rely on exogenous increases in credit supply, we similarly exploit the randomized experiment conducted by the e-wallet provider as an exogenous shock to credit supply to examine the impact of BNPL on consumer spending. We then return to our transaction data to explore the dynamics of indebtedness and BNPL usage.

BNPL has several unique features making it well-suited for studying this issue, as it may have different impacts on consumption and indebtedness compared with traditional credit lines increases on credit cards. First, the BNPL in the e-wallet we study targets young consumers with low income and no credit card, who are most likely to be financially constrained. As shown in Section 4, BNPL achieves higher credit coverage for consumers than credit cards do. In our sample, 68.31% of consumers lack credit card access for online transactions and 70.43% for on-site, while 78% and 84.14% have access to BNPL for online and on-site transactions, respectively. Second, the BNPL service in e-wallet credit provides more options and flexibility for consumers and is particularly attractive for those who are uncomfortable carrying credit card balances but comfortable borrowing for specific purchases. These features of BNPL could have profound impacts on consumer spending. Moreover, as demonstrated in Section 3, BNPL is frequently used for low-value purchases as the new cash in e-wallet transactions, which may lead to different results on household indebtedness relative to the results from overuse of credit cards.

5.1 BNPL and Consumer Spending

To formally check the impact of BNPL on consumer spending, we exploit a randomized experiment conducted by the e-wallet provider we study. As the sample here is only a subsample of the randomized experiment, coupled with the intention to improve the precision of the estimate, we employ the difference-in-differences approach to examine the impact of the introduction of BNPL on consumer spending,

thanks to the fact that our dataset covers both the pre-treatment and post-treatment periods from January to July 2017. Specifically, we are comparing the outcomes of the treated users after gaining access to BNPL relative to the time before they had credit access and relative to users in the control group. The estimation model is specified as follows:

$$Consumption_{it} = \beta_0 + \beta_1 treat_i * post_t + \gamma \ln \ cashflow_{it} + \lambda_i + \delta_t + \varepsilon_{it}$$
(3)

Here, *Consumption*_{it} is the amount of spending through the e-wallet for consumer *i* in month *t*; *treat*_i is a dummy variable that equals 1 for treated users and 0 for untreated users; *post*_t is a dummy variable that equals 1 for user-month observations in June and July during the two-month experiment and 0 otherwise. The control variable $ln_cashflow_{it}$ is the logarithm of e-wallet cash flows in the last year to proxy for users' financial position. We also include both user fixed effects (λ_i) and month fixed effects (δ_t) in the regression to control for the impacts of unobserved time-invariant user characteristics and macroeconomic factors.³⁴ ε_{it} is the error term. We expect a significant and positive β_1 , which would indicate a positive spending response to the introduction of BNPL.

The aggregate consumption boosting effect in the full sample is reported in Table 13 Column (1); the heterogeneous effects of BNPL on single-access users and dual-access users can be seen in Columns (2) and (3). Moreover, in addition to aggregate consumption, it is unclear whether BNPL access leads to more or less credit card consumption overall. Column (4) focuses on the credit card consumption response to BNPL access among dual-access users.

In all columns, the interaction term between the treatment and the time indicator, $treat_i * post_i$, is positive and statistically significant, implying that the introduction of BNPL increases consumer spending. In the full sample, compared with the average monthly consumption of untreated users, consumption of the treated group increases by

³⁴ Since user fixed effects are included in the regression, we do not report the coefficient estimates on users' age and gender. These two variables are fully absorbed by user fixed effects.

RMB ¥86, corresponding to approximately 5.62% of total average monthly consumption (0.086/1.53) as reported by the National Bureau of Statistics of China.³⁵ For single-access users in Column (2), the average monthly consumption increases by RMB ¥85 after gaining access to BNPL; for dual-access users in Column (3), the average consumption increases by RMB ¥204.³⁶

In Column (4), credit card consumption also increases for treated dual-access users, consistent with findings in Section 3 that BNPL access reduces credit card usage in transactions but do not completely substitute credit cards in the big-ticket purchases, corroborating our claim that BNPL plays the role as digital cash rather than credit cards.

| | Dependent variable: consumption (monthly, thousand RMB) | | | | |
|---------------------|---|---------------|-------------|--------------|--|
| | (1) | (2) | (4) | | |
| | Full sample | Single-access | Dual-access | Dual-access | |
| | | | | (credit card | |
| | | | | consumption) | |
| Treat×Post | 0.086*** | 0.085*** | 0.204*** | 0.038*** | |
| | (0.002) | (0.002) | (0.047) | (0.009) | |
| Ln_cashflow | 0.014*** | 0.014*** | 0.044*** | -0.0008 | |
| | (0.0003) | (0.0003) | (0.005) | (0.001) | |
| User fixed effects | YES | YES | YES | YES | |
| Month fixed effects | YES | YES | YES | YES | |
| Observations | 700,000 | 695,149 | 4,851 | 4,851 | |
| \mathbb{R}^2 | 0.225 | 0.226 | 0.199 | 0.276 | |

Table 13. The impact of BNPL on consumer spending

Notes: ***p < 0.01; **p < 0.05; *p < 0.1.

5.2 BNPL, Default, and Consumer Indebtedness

Since the introduction of BNPL boosts consumer spending significantly, some may argue that BNPL induces overspending and increases consumer indebtedness or

³⁵ Average yearly consumption by Chinese people was RMB ¥18,322 in 2017. See

http://www.stats.gov.cn/tjsj/zxfb/201801/t20180118_1574931.html; accessed on August 5, 2020.

³⁶ The higher effect for dual-access users may be attributed to their higher average wealth. This coefficient estimation for dual-access users should be interpreted with caution due to its limited sample size and much higher standard error. The 95% confidence interval of *Treat*×*Post* in Column (3) is (0.1124, 0.2952). In comparison, the 95% confidence interval of the coefficient estimate for *Treat*×*Post* in Column (1) is (0.0818, 0.0903). Though unreported, we also explored the effect of BNPL access on BNPL consumption. The coefficients on *treat_i*post_i* are 0.029 for single-access users and 0.054 for dual-access users.

default, especially for single-access users who are underserved by banks but have now received consumer credit for the first time.

We first compare single-access and dual-access users in Table 14 concerning the BNPL usage as a means of payment and repay the balance each month (transactors) or the usage as a source of credit and incur interest charges for unpaid debt (revolvers). Note that the overall default rate is below 1% in our sample and below 2% among all BNPL users in the population.³⁷ But revolvers are common. About two-thirds of single-access consumers use BNPL as a means of payment; the fraction of revolvers is 32.33%, which is significantly lower than the fraction for dual-access users (40.78%). In addition, the ratio of unpaid BNPL credit to used credit balance is 28.13% for single-access revolvers, which is also lower than that for dual-access revolvers (31.48%). Moreover, single-access users choose to pay with BNPL for 58.24% of their transactions, while dual-access users use BNPL for 56.97% of their transactions.

| | All e-wallet | Dual-access | Single-access | |
|-----------------------------|------------------|-------------|---------------|----------|
| | credit users (1) | users (2) | users (3) | (2)–(3) |
| The fraction of revolvers % | 35.16 | 40.78 | 32.33 | 8.45*** |
| Unpaid BNPL debt/used | | | | |
| credit % | 29.32 | 31.48 | 28.13 | 3.35*** |
| BNPL transactions/all | | | | |
| transactions % | 57.81 | 56.97 | 58.24 | -1.27*** |

Table 14. BNPL usage for different purposes between single-access users and dual-access users

Notes: 1. The last column shows the difference between columns (2) and (3). 2. ***p < 0.01; **p < 0.05; *p < 0.1.

Taken together, while single-access users appear to use BNPL more frequently, they use the credit more cautiously, demonstrated by a lower proportion of revolvers and a lower unpaid-debt ratio for revolvers.³⁸ This behavior is voluntary as the e-wallet provider benefits from receiving interest payments and would not require BNPL users to reduce usage or impose penalty without default. To some extent, this

³⁷ Data source: the 2020 prospectus from the data provider. The e-wallet provider also carefully manages default risk: in addition to the diversification of borrowers across demographic groups, the provider reduces a user's credit score within the e-commerce ecosystem if a user defaults, and utilizes proprietary data to screen for eligible and trusted borrowers.

³⁸ The overall debt level of dual-access users is inferred to be higher than that of single-access users because of the higher BNPL debt ratio. Given our findings in Table 13 that BNPL access increases credit card consumption, it is unlikely that dual-access users would engage in debt restructuring by turning credit card debt into BNPL debt.

alleviates our concern that single-access users who receive consumer credit for the first time may overuse BNPL due to the oft-cited lack of financial literacy and budgeting education.

We return to the transaction-level payment data and analyze how consumers change their usage of BNPL when they have unpaid debt and incur interest charges. It is important to consider the impact on the usage of BNPL as the payment method when an e-wallet user faces interest charges. If users choose to use more BNPL after incurring interest, they are likely to develop into deep revolvers bearing a heavier debt burden and a default can hurt their credit building effort within the e-commerce system. In contrast, if e-wallet users reduce their BNPL usage once faced with interest expenses, then they are temporarily revolvers and likely to become transactors again in the future, alleviating the concern of excessive indebtedness.

Here we examine the relationship between BNPL usage and interest expenses. We have four proxies to measure the extent to which a user incurs unpaid debt and interest charges in e-wallet credit: (1) a dummy variable indicating whether a user has unpaid debt (*D_Unpaid_BNPL*); (2) the logarithm of one plus the unpaid debt balance (*Ln_unpaid_BNPL_debt*); (3) the fraction of unpaid debt in used e-wallet credit (*Unpaid_BNPL_debt*); (4) three dummy variables indicating whether the unpaid BNPL debt belongs to low-level, medium-level, or high-level groups (*Unpaid_BNPL_debt_Low/Middle/High*). We separately include these four proxies in Equations (1) and (2), which examine the determinants of the usage of BNPL as the payment instrument in on-site and online transactions. To differentiate the behavior of single-access users and that of dual-access users, we also include the interaction terms between the first three proxies and the dummy variable indicating whether a user has access to a credit card (*D_Consumer_CC*).

Table 15. The impact of incurring interest on the usage of BNPL as the payment instrument in

on-site transactions

| | The dependent variable: a dummy variable indicating whether BNPL is used in an on-site transaction | | | |
|--|--|---------------------------------------|-----------------|---------------------|
| | (1) | $\frac{PL \text{ is used in a}}{(2)}$ | | |
| D_Consumer_CC ^d | 0.039*** | 0.043*** | (3) 0.026*** | (4) |
| D_consumer_cc | (0.000) | (0.000) | (0.000) | |
| D Marchant DNDI | (0.000) | (0.000) | (0.000) | (0.000) 0.034*** |
| D_Merchant_BNPL ^d | | | | |
| D. Marshaut, CCd | (0.000) -0.071*** | (0.000) -0.070*** | (0.000) | (0.000) |
| D_Merchant_CC ^d | | | -0.060*** | -0.074*** |
| | (0.000) | (0.000) | (0.000) | (0.000) |
| Consumer_preference_BNPL ^d | 0.391*** | 0.383*** | 0.270*** | 0.382*** |
| D_Unpaid_BNPL ^d | (0.000) -0.060*** | (0.000) | (0.000) | (0.000) |
| | (0.000) | | | |
| D_Unpaid_BNPL ^d ×D_Consumer_CC ^d | -0.010*** | | | |
| | (0.000) | | | |
| Ln_unpaid_BNPL_debt | (0.000) | -0.008*** | | |
| | | (0.000) | | |
| Ln_unpaid_BNPL_debt×D_Consumer_CC ^d | | (0.000) -0.002^{***} | | |
| | | (0.000) | | |
| Unpaid_BNPL_debt % | | (0.000) | -0.184*** | |
| | | | (0.000) | |
| Unpaid_BNPL_debt %× | | | -0.006* | |
| D_Consumer_CC ^d | | | (0.077) | |
| D_consumer_cc | | | (0.077) | |
| Unpaid_BNPL_debt_Low ^d | | | | 0.034*** |
| - | | | | (0.000) |
| Unpaid_BNPL_debt_Middle ^d | | | | -0.068*** |
| · | | | | (0.000) |
| Unpaid_BNPL_debt_High ^d | | | | -0.172*** |
| 1 0 | | | | (0.000) |
| Control variables | YES | YES | YES | YES |
| Industry FE | YES | YES | YES | YES |
| Day FE | YES | YES | YES | YES |
| Observations | 448,270 | 442,493 | 393,482 | 448,270 |
| Pseudo-R ² | 0.13 | 0.13 | 0.11 | 0.14 |

Notes: 1. Unpaid_BNPL_debt %=unpaid debt / used e-wallet credit. 2. The superscript d indicates a dummy variable. 3. *** p < 0.01; ** p < 0.05; * p < 0.1. 4. We report the marginal effects for the logit model.

Table 16. The impact of incurring interest on the usage of BNPL as the payment instrument in

online transactions

| | Dependent variable: a dummy variable indicating | | | |
|--|---|-----------|-----------|-----------|
| | whether BNPL is used in an online transaction | | | |
| | (1) | (2) | (3) | (4) |
| D. C | -0.075*** | -0.053*** | -0.077*** | -0.055*** |
| D_Consumer_CC ^d | (0.000) | (0.000) | (0.000) | (0.000) |
| | 0.376*** | 0.383*** | 0.276*** | 0.374*** |
| Consumer_preference_BNPL ^d | (0.000) | (0.000) | (0.000) | (0.000) |
| D_Unpaid_BNPL ^d | -0.026*** | | | |
| | (0.000) | | | |
| D_Unpaid_BNPL ^d ×D_Consumer_CC ^d | 0.041*** | | | |
| | (0.000) | | | |
| Ln_Unpaid_BNPL_debt | | -0.003*** | | |
| | | (0.000) | | |
| Ln_Unpaid_BNPL_debt×D_Consumer_CC ^d | | 0.002*** | | |
| | | (0.000) | | |
| Unpaid_BNPL_debt % | | | -0.180*** | |
| | | | (0.000) | |
| Unpaid_BNPL_debt %×D_Consumer_CC ^d | | | 0.057*** | |
| | | | (0.000) | |
| Unpaid_BNPL_debt_Low ^d | | | | 0.083*** |
| 1 | | | | (0.000) |
| Unpaid_BNPL_debt_Middle ^d | | | | -0.027*** |
| | | | | (0.000) |
| Unpaid_BNPL_debt_High ^d | | | | -0.124*** |
| | | | | (0.000) |
| Control variables | YES | YES | YES | YES |
| Industry FE | YES | YES | YES | YES |
| Day FE | YES | YES | YES | YES |
| Observations | 433,136 | 425,028 | 352,403 | 433,136 |
| Pseudo-R ² | 0.10 | 0.09 | 0.07 | 0.11 |

Notes: 1. Unpaid_BNPL_debt %=unpaid debt / used credit e-wallet credit. 2. The superscript *d* indicates a dummy variable. 3. *** p < 0.01; ** p < 0.05; * p < 0.1. 4. We report the marginal effects for the logit model.

Tables 15 and 16 present the regression results for on-site transactions and online transactions, respectively. The coefficients on the first three proxies are all negative and statistically significant, suggesting that single-access users reduce their usage of BNPL as the payment tool in both on-site and online transactions when they incur unpaid debt and interest charges. Relative to single-access users, dual-access users reduce their usage of BNPL as a payment tool in on-site transactions to a greater extent, while they reduce their usage of BNPL to a lesser extent in online transactions.

According to the results in Columns (4) in Tables 15 and 16, whilst BNPL users exhibit a slight increase in BNPL usage when faced with low-level unpaid debt, potentially to address their liquidity needs, their usage of BNPL significantly decreases as their debt levels enter the middle or high terciles. This could be intuitively attributed to the higher delinquent interest rate charged by BNPL providers compared to credit cards.³⁹

Together the results show that single-access users are careful in using BNPL even when they receive consumer credit for the first time. They have a lower fraction of revolvers and a lower ratio of unpaid debt to used e-wallet credit; they also voluntarily reduce their usage of BNPL as the payment tool in both on-site and online transactions after they incur unpaid debt and interest charges. Our findings are in line with those of Agarwal et al. (2021), who exploited a large-scale microcredit program in Rwanda. They found that a sizable share of first-time borrowers, who demonstrated lower default risks, successfully switched to bank credit later. The prudence of the first-time borrowers from non-bank credit sources could boost their credit scores and pave a pathway to future bank credit.

5.3 A Different BNPL Model

While it is difficult to make welfare judgement on whether the boosted consumption is overspending or not, our results suggest that the introduction of BNPL may not necessarily lead to additional default and indebtedness. This is in sharp contrast to existing literature documenting how BNPL leads to more defaults and delinquencies. Several salient features of the Chinese model of BNPL may explain the differences.

First, although BNPL eligibility is very inclusive, many digital wallets still screen carefully ex ante. E-wallets embedded in a large e-commerce ecosystem have access to real-time, high-frequency transaction data of borrowers (micro, small, and medium-sized enterprises operating on the platform), as well as online financial and

³⁹ This is consistent with the common critique that BNPL is "inclusive but expensive" relative to credit cards. See https://m.21jingji.com/article/20201102/herald/62dfca7696f0148a9353a88dbd9eedf5.html.

behavioral data (e.g., gross merchandise volume, authenticity/illegal sales, customer ratings, credit card payments, online shopping payments, fund transfers, utility payments, etc.). The data-network-activity (DNA) feedback loop helps BigTech firms to identify the characteristics of their clients (Gambacorta, et al., 2023), and the e-wallet providers can utilize and develop a proprietary credit-reference system to screen eligible consumers effectively while still remaining more inclusive than banks (Chen, Huang, Lin, and Sheng, 2022). In addition, BNPL providers in China apply big data analytics on both traditional data and proprietary information such as digital footprints, and use machine learning models to improve the accuracy of loan default prediction, all of which help reduce defaults and delinquencies ex ante.

Second, BNPL users also have incentives to avoid default and excessive indebtedness. The e-wallet provider can monitor borrower behavior in the ecosystem and more effectively enforce penalties. For example, it could temporarily suspend access to certain advanced services on the platform, and it might deduct balances from their digital wallets for repayment purposes (Chen, Huang, Lin, and Sheng, 2022; Gambacorta et al., 2023). This is particularly useful in reducing borrower moral hazard because of the large network effects in the super app and the high cost of switching between e-commerce platforms. Consumers also care about improving their credit score within the super app ecosystem, which allows users to access more features and services. A parallel in the U.S. is PayPal Credit, where users with higher credit scores access more platform features, such as lower fees and special offers. This motivates individuals to maintain and improve creditworthiness, strengthening their commitment to responsible borrowing and repayment.

Third, culture and tradition can play a role too. Tajaddini and Gholipour (2017) investigate the relationship between national cultural characteristics and default on mortgages (DOM), documenting that borrowers from countries with high individualism default more on their mortgages. Among the 42 countries examined from 2010 to 2013, China has low individualism and the lowest mortgage default rate. Recently, Zhang (2022) uses data from a large peer-to-peer lending platform to

document that borrowers from regions with stronger moral norms have lower loan default probabilities.

Obviously, the first two explanations have external validity because BNPL providers in other countries are converging to the super app model in China and are using similar ex ante screening or ex post monitoring/enforcement/incentive provisions. Even the third may apply to other developing economies sharing similar culture or tradition with China. Finally, we note the limited time span of the experiment. It could also be that individuals temporarily increase total consumption expenditure using BNPL credit without reducing consumption using credit cards or increasing indebtedness, all due to mental accounting (Gelman and Roussanov, 2022), which is a behavioral pattern not limited to Chinese consumers.

6. Conclusion

Digital wallets and Buy-Now-Pay-Later (BNPL) have seen exponential growth around the globe, especially during the COVID-19 pandemic. Yet, their inner workings, the link between digital payments and consumer credit, and the effect of BNPL on consumer behavior are little understood, especially in economies transitioning from being cash-heavy to cashless, which have seen the largest market growth and potential.

We conduct the first investigation into the hitherto black box of e-wallets to better understand digital payment competition, FinTech credit provision to households, and their implications for consumer spending and debt management. Internal payment options in e-wallets dominate external options, with BNPL crowding out other payment choices while complementing credit cards for small-value purchases. The e-wallet in our study has a pronounced credit-expansion effect for underserved consumers through BNPL adoption by both consumers and merchants, and in less developed regions and for individuals underserved by banks.

Utilizing the large-scale, matched BNPL transactions that we observe directly for

the first time in the literature, as well as the randomized control experiment conducted by the e-wallet provider, we also find that BNPL significantly boosts consumption; yet, consumers—especially those who rely solely on BNPL for credit access—are careful with overspending and indebtedness. Contrasting with studies on BNPL using bank data or surveys from developed countries where BNPL providers often started with pure-play merchant partner models, our paper highlights that introducing BNPL through super apps with proper screening and well-designed incentives can lead to opposite outcomes in consumer indebtedness and loan defaults.⁴⁰ Overall, our findings suggest that e-wallet credit in the form of BNPL is becoming a new form of cash and seems to facilitate financial inclusion.

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⁴⁰ Arguably, after 2020, regulatory changes in China's FinTech consumer credit services, such as tightened requirements for BNPL providers to report credit information to the central bank's credit reference system and reduction in credit limits for young adults, especially college students, further reinforce our findings regarding the with BNPL consumer caution usage and debt management. For more details, see https://www.wsj.com/articles/jack-mas-ant-group-slashes-credit-limits-for-some-younger-borrowers-in-china-1160 8738467?mod=article inline and

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