# Buy Now Pay Later Credit: User characteristics and effects on spending patterns 

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## Background and Motivation

Consumers have substantial demand for short term unsecured credit

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Not revolving lines of credit but are instead installment loans with a down-payment
which is due at the point of sale and a fixed repayment schedule
Offered through retailers and tied to price of the product
Offered with easy lending terms i.e. no credit checks, zero interest, no credit bureau
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## "Pay-in-4" is the most popular BNPL product

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How does "Pay-in-4" work?
    25% of the purchase price is required at the time of purchase
    additional equal payments are required every 2 weeks thereafter.
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In this paper we provide a first look into BNPL market with two objectives:

1. We use transactions level bank account and credit card data to provide insights about BNPL use.
2. We study consumer responses to new Fintech credit products, adding to the existing literature studying consumer spending responses to income and liquidity.

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An incomplete understanding of contract terms
Reference dependent preferences.

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## Background

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| BNPL | Credit | Pay-in-4 | Late | Non Pay-in-4 <br> Option? <br> Option? | Interest <br> Rates | Negative <br> Reporting | Positive <br> Reporting |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Check | Opfirm | Yes - Soft | Yes | None | Yes | $0-30 \%$ | Yes | Yes

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| Affirm | Yes - Soft | Yes | None | Yes | $0-30 \%$ | Yes | Yes |
| Afterpay | No | Yes | $\$ 8 / \max 25 \%$ | No | NA | No | No |
| Klarna | Yes - Soft | Yes | $\max 25 \%$ | Yes | $0-20 \%$ | Yes | No |
| Quadpay | Yes - Soft | Yes | $\$ 7 / \max \$ 21$ | No | NA | Yes | No |
| Sezzle | Yes - Soft | Yes | $\$ 10$ | No | NA | Yes | Yes |

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BNPL availability has grown exponentially since 2020

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Integration with top 10k Websites


## Data \& Patterns of BNPL Use

## Transactions level data

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We are able to observe BNPL use through bank account and credit card transactions
level data for around }10\mathrm{ million active users in the US
We identify BNPL transactions by making use of merchant classification provided by
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## Population summary statistics by BNPL use/year

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|  | As of December |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 2017 |  | 2018 |  | 2019 |  | 2020 |  |
|  | BNPL user |  | BNPL user |  | BNPL user |  | BNPL user |  |
|  | No | Yes | No | Yes | No | Yes | No | Yes |
| Mean, \% |  |  |  |  |  |  |  |  |
| Renter | 8.7 | 12.3 | 8.9 | 13.1 | 9.4 | 13.8 | 10.0 | 15.2 |
| Credit card user | 43.4 | 37.6 | 42.5 | 35.8 | 43.8 | 35.1 | 39.4 | 32.7 |
| Active saver | 10.6 | 9.4 | 10.3 | 11.1 | 9.6 | 11.7 | 6.3 | 7.9 |
| \$400 buffer | 68.6 | 67.3 | 69.6 | 67.6 | 68.9 | 65.8 | 73.9 | 74.2 |
| Paid overdrafts | 4.9 | 7.6 | 4.3 | 8.0 | 4.0 | 9.2 | 3.0 | 7.7 |
| Median, \$ |  |  |  |  |  |  |  |  |
| Salary | 3,639 | 3,687 | 3,534 | 3,659 | 3,676 | 3,736 | 4,355 | 4,512 |
| Essential spending | 443 | 747 | 462 | 795 | 509 | 849 | 486 | 861 |
| Discretionary spending | 277 | 498 | 284 | 531 | 320 | 559 | 282 | 549 |
| Bills | 261 | 387 | 273 | 415 | 291 | 437 | 287 | 442 |
| Retail spending | 202 | 435 | 206 | 462 | 254 | 529 | 278 | 611 |
| Credit transactions | 5,428 | 5,936 | 5,400 | 6,021 | 5,801 | 6,472 | 6,741 | 7,878 |
| Debit transactions | 5,287 | 5,951 | 5,598 | 6,355 | 6,142 | 6,922 | 6,312 | 7,500 |
| Sample size | 254,018 | 28,008 | 251,162 | 30,933 | 227,763 | 32,272 | 178,961 | 29,719 |

## User summary by providers

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|  | As of December 2019 |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  | Affirm | Afterpay | Klarna | Quadpay | Sezzle |
| Mean, \% | 14 | 14.6 | 14.1 | 16.8 | 15.1 |
| Renter | 40.1 | 28.3 | 32.2 | 25.6 | 28.7 |
| Credit card user | 11.8 | 12.8 | 11.2 | 10.7 | 10.8 |
| Active saver | 71.8 | 60.4 | 62.8 | 57.9 | 60.9 |
| \$400 buffer | 7.9 | 11.7 | 10.6 | 12.8 | 11.1 |
| Paid overdrafts |  |  |  |  |  |
| Median, \$ | 4,199 | 3,311 | 3,524 | 3,211 | 3,421 |
| Salary | 908 | 836 | 859 | 815 | 907 |
| Essential spending | 602 | 574 | 594 | 565 | 573 |
| Discretionary spending | 475 | 428 | 439 | 433 | 449 |
| Bills | 562 | 547 | 547 | 555 | 644 |
| Retail spending | 7,447 | 5,817 | 6,193 | 5,670 | 5,947 |
| Credit transactions | 7,929 | 6,241 | 6,593 | 6,076 | 6,326 |
| Debit transactions | 16,745 | 14,570 | 10,577 | 2,242 | 4,628 |
| Users (out of 260,035) |  |  |  |  |  |

## Decision tree analysis highlighting User characteristics associated with BNPL use



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BNPL users tend to:
Spend more on non-essential goods
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## Decision tree analysis highlighting User characteristics associated with BNPL use

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## BNPL usage has grown exponentially since 2020

Panel C: Underlying users BNPL/CC


## BNPL usage has grown exponentially since 2020

BNPL adoption by provider: Affirm


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BNPL adoption by provider: All


## Most people use BNPL only a few times

Most people use BNPL only a few times


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## People use BNPL more during the holiday period

Frequency of BNPL Down-payments by Week of the Year


## What should we expect?

In a traditional model, access to zero-interest unsecured credit has two immediate effects

1. Consumers will optimally increase current consumption via intermporal substitution effects and a reduction in precautionary savings motive Consumers will have a greater capacity to smooth consumption across liquiity shocks.
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5. Consumers will have a greater capacity to smooth consumption across liquiity shocks.

We investigate both effects in detail.

## BNPL access and spending

## Within user difference in differences analysis

> We begin our analysis by documenting spending responses to first time BNPL use Specifically we run regressions of the following form at the calendar week level:

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y_{i t}=\alpha_{i t}+\sum_{k=-12}^{24} \gamma_{k} \mathbb{1}\{\text { First_BNPL_i }-t=k\} \times \text { Tit }+\varepsilon_{i t}
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> Total spending increases, driven mainly by retail spending but also spending in other categories

Total spending increases, driven mainly by retail spending but also spending in other categories

Panel A: Total spending


Total spending increases, driven mainly by retail spending but also spending in other categories

Panel B: Total spending without BNPL


Total spending increases, driven mainly by retail spending but also spending in other categories

Panel C: Retail spending


Total spending increases, driven mainly by retail spending but also spending in other categories

Panel D: Essential spending


Total spending increases, driven mainly by retail spending but also spending in other categories

Panel E: Discretionary spending


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Panel F: Other spending


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Panel C: Retail spending


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BNPL use shifts the allocation of expenditure towards retail goods

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Panel A: Retail spending


## BNPL use shifts the allocation of expenditure towards retail goods



## BNPL use shifts the allocation of expenditure towards retail goods

Panel C: Discretionary spending


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BNPL use is associated with a reduction in liquidity

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Panel A: Bank account balance


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Panel B: Overdraft


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Panel C: Low balance fees


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Panel A: Bank account balance


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Panel B: Overdraft


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The within-user analysis is consistent with an increase in spending as a result of BNPL use, however...

- The timing of BNPL use might be correlated with unobserved time-varying user-specific expenditure trends
- To make progress and isolate causal effects, we construct an instrument for BNPL access.


## We make use of heterogeneity in the timing of BNPL adoption by retailers



We make use of heterogeneity in the timing of BNPL adoption by retailers
Last Year ( $T-1$ )


We make use of heterogeneity in the timing of BNPL adoption by retailers

$$
\text { Last Year }(T-1)
$$

Target = 33\%
Sam's Club = 8\%


This Year ( $T$ )


We use this binary exposure instrument to estimate coefficients in the structural equation:
variable equal to one after the first time a person uses BNPL equation:

$$
y_{i t}=\alpha_{i t}+\beta \text { Post }_{i t}+\varepsilon_{i t}
$$

i.e. we use $E_{i, t}$, a binary exposure variable, to instrument for Post ${ }_{i t}$, which is an indicator variable equal to one after the first time a person uses BNPL

# Examples of major retailers identified as offering BNPL in our data 



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| Retailer | Customers <br> $\%$ | Revenue <br> bn. \$ |
| :--- | :---: | :---: |
| Target | 49.1 | 52.6 |
| Bed Bath \& Beyond | 19.1 | 5.0 |
| Michaels | 18.8 | 2.7 |
| Sam's Club | 17.2 | 22.4 |
| GameStop | 12.7 | 2.6 |
| IKEA | 11.6 | 4.5 |
| Nordstrom | 11.5 | 15.0 |
| Etsy | 10.7 | 3.0 |
| Forever 21 | 10.5 | 1.2 |
| Sephora | 9.2 | 2.5 |

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| Sephora | 9.2 | 2.5 |

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## Post-BNPL availability, total spending increases

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| Total Spend |  |  |  |
| :--- | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS |
| Post | $40.16^{* * *}$ <br> $(1.39)$ | $60.47^{* * *}$ <br> Exposure | 7.512*** <br> $(1.479)$ |
| KP Wald F Stat |  |  | 1,163 |

## Post-BNPL availability, total spending increases

| Retail Spend |  |  |  |
| :--- | :---: | :---: | :--- |
|  | Fixed Effects | Reduced Form | TSLS |
| Post | $20.16^{* * *}$ |  | $53.58^{* * *}$ <br> $(0.37)$ |
| Exposure |  | 6.758*** <br> $(0.449)$ |  |
| KP Wald F Stat |  |  | 1,220 |

## Post-BNPL availability, total spending increases for non CC users

Total Spend

|  | Credit Card User |  |  | Not a Credit Card User |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS | Fixed Effects | Reduced Form | TSLS |
| Post | $\begin{aligned} & 31.81 * * * \\ & (1.512) \end{aligned}$ |  | $\begin{gathered} -1.844 \\ (16.68) \end{gathered}$ | $\begin{aligned} & 41.52^{* * *} \\ & (1.567) \end{aligned}$ |  | $\begin{aligned} & 68.26^{* * *} \\ & (10.57) \end{aligned}$ |
| Exposure |  | $\begin{array}{r} -0.168 \\ (1.517) \end{array}$ |  |  | $\begin{aligned} & 8.869^{* * *} \\ & (1.533) \end{aligned}$ |  |
| KP Wald F Stat |  |  | 690.4 |  |  | 959.3 |

## Post-BNPL availability, total spending increases for non CC users

Retail Spend

|  | Credit Card User |  |  | Not a Credit Card User |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS | Fixed Effects | Reduced Form | TSLS |
| Post | $\begin{aligned} & 18.75^{* * *} \\ & (0.391) \end{aligned}$ |  | $\begin{aligned} & 42.26^{* * *} \\ & (4.132) \end{aligned}$ | $\begin{aligned} & 20.45 * * * \\ & (0.394) \end{aligned}$ |  | $\begin{aligned} & 57.08^{* * *} \\ & (2.650) \end{aligned}$ |
| Exposure |  | $\begin{aligned} & 3.943 * * * \\ & (0.443) \end{aligned}$ |  |  | $\begin{aligned} & 7.468^{* * *} \\ & (0.455) \end{aligned}$ |  |
| KP Wald F Stat |  |  | 734 |  |  | 991 |

Post-BNPL availability, the consumption basket shifts towards retail spending and away from other discretionary type spending

Post-BNPL availability, the consumption basket shifts towards retail spending and away from other discretionary type spending

| Retail Spend/Total |  |  |  |
| :--- | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS |
| Post | $0.0335^{* * *}$ |  | $0.0629^{* * *}$ <br> Exposure |
|  |  | $0.0005)$ <br> $(0.0065)$ |  |
| KP Wald F Stat |  |  | 1,165 |

Post-BNPL availability, the consumption basket shifts towards retail spending and away from other discretionary type spending for cc users as well

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Retail Spend/Total

|  | Credit Card User |  |  | Not a Credit Card User |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS | Fixed Effects | Reduced Form | TSLS |
| Post | $\begin{gathered} 0.0290 * * * \\ (0.000535) \end{gathered}$ |  | $\begin{gathered} 0.0447 * * * \\ (0.00578) \end{gathered}$ | $\begin{gathered} 0.0369 * * * \\ (0.000563) \end{gathered}$ |  | $\begin{aligned} & \hline 0.0720 * * * \\ & (0.00430) \end{aligned}$ |
| Exposure |  | $\begin{aligned} & 0.00381 * * * \\ & (0.000517) \end{aligned}$ |  |  | $\begin{aligned} & 0.00806^{* * *} \\ & (0.000531) \end{aligned}$ |  |
| KP Wald F Stat |  |  | 680.1 |  |  | 892.7 |

Post-BNPL availability, liquidity declines and unsecured borrowing increases

Post-BNPL availability, liquidity declines and unsecured borrowing increases

## Post-BNPL availability, liquidity declines and unsecured borrowing increases

| Balance Estimate |  |  |  |
| :--- | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS |
| Post | $-659.4^{\star * *}$ |  | $-997.2^{* *}$ |
| Exposure | $(63.9)$ |  | $(408.7)$ |
|  |  | $-123.9^{* *}$ <br> $(51.4)$ |  |
| KP Wald F Stat |  |  | 1,163 |

## Post-BNPL availability, liquidity declines and unsecured borrowing increases

| Overdraft Fee |  |  |
| :--- | :---: | :---: |
|  | Fixed Effects | Reduced Form |
| Post | $0.00263^{* * *}$ |  |
| Exposure | $(0.00026)$ |  |
|  |  | $0.000588^{* * *}$ |
|  |  | $(0.000203)$ |
| KP Wald F Stat |  |  |

Post-BNPL availability, liquidity declines and unsecured borrowing increases - for non CC users

Post-BNPL availability, liquidity declines and unsecured borrowing increases - for non CC users

Balance Estimate

|  | Credit Card User |  |  | Not a Credit Card User |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS | Fixed Effects | Reduced Form | TSLS |
| Post | $\begin{gathered} -833.1 * * * \\ (110.4) \end{gathered}$ |  | $\begin{array}{r} -894.2 \\ (1.136) \end{array}$ | $\begin{gathered} -462.9^{* * *} \\ (54.49) \end{gathered}$ |  | $\begin{aligned} & -1.663^{* * *} \\ & (360.9) \end{aligned}$ |
| Exposure |  | $\begin{aligned} & -81.46 \\ & (103.5) \end{aligned}$ |  |  | $\begin{gathered} -216.0^{* * *} \\ (47.53) \end{gathered}$ |  |
| KP Wald F Stat |  |  | 690.4 |  |  | 959.3 |

## Post-BNPL availability, liquidity declines and unsecured borrowing increases - for

 non CC usersOverdraft Fee

|  | Credit Card User |  |  | Not a Credit Card User |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fixed Effects | Reduced Form | TSLS | Fixed Effects | Reduced Form | TSLS |
| Post | $\begin{gathered} 0.00157^{* * *} \\ (0.000242) \end{gathered}$ |  | $\begin{gathered} -0.00123 \\ (0.00227) \end{gathered}$ | $\begin{gathered} 0.00269^{* * *} \\ (0.000361) \end{gathered}$ |  | $\begin{aligned} & 0.00458 * * \\ & (0.00232) \end{aligned}$ |
| Exposure |  | $\begin{gathered} -0.000112 \\ (0.000207) \end{gathered}$ |  |  | $\begin{gathered} 0.000595^{*} \\ (0.000304) \end{gathered}$ |  |
| KP Wald F Stat |  |  | 690.4 |  |  | 959.3 |

# BNPL access and consumption smoothing 

The relationship between spending and weekly income flattens especially for lower income consumers

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BNPL spending increases substantially in periods when weekly salary drops substantially

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# Taking stock of our findings 

```
We find that BNPL access is associated with:
    Increased spending
    2. A flatter rolationshin between spending and income
    3. A shift in spending towards retail goods (for all users)
    4. An reduction in liquid buffers (for those with less access to liquidity ex ante)
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The "liquidity fly paper effect"

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BNPL disproportionately provides the additional liquidity "sticks" in retail liquidity for retail purchases

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BNPL disproportionately provides liquidity for retail purchases
the additional liquidity "sticks" in retail and leads to higher retail consumption

## Our findings are also consistent with a "liquidity flypaper effect"

> Ad-hoc budgeting rules or mental accounting combined with consumer myopia i.e consumers budget a certain amount for, say, clothing expenditure in each period 2. consumers fail to fully recognize how future payments will impact future liquidity.

## Our findings are also consistent with a "liquidity flypaper effect"

$$
\begin{aligned}
& \text { How? } \\
& \text { Ad-hoc budgeting rules or mental accounting combined with consumer myopia i.e. } \\
& \text { 1. consumers budget a certain amount for, say, clothing expenditure in each period. } \\
& \text { 2. consumers fail to fully recognize how future payments will impact future liquidity. }
\end{aligned}
$$

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```
Do people "code" BNPL repayments like other bill payments?
We can identify type of retailer, and check to see if spending increases are confined
to more granular retail buckets as confirmation of mental accounting type forces at
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# Work in progress to pin down the "liquidity flypaper effect" 

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```
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3. We find that BNPL use is associated with:
4. We argue that these findings are consistent with increased credit access causing a reduction in the precautionary savings motive and also a "liquidity flypaper effect"
5. We find that BNPL use is associated with:

Increased spending
smoother spending relative to income
increased allocation towards retail goods
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## Wrap up

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- Increased retail liquidity leads to increased retail spending

```
    Mis-coding of repayments as retail spending can create a persistent increase in retail
    allocation.
    This additional spending is financed with existing liquidity
    These effects are also true for people who do not face binding liquidity constraints.
```


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[^0]:    In this paper we provide a first look into BNPL market with two objectives:

    1. We use transactions level' bank account and cred't card data to provide insights about BNPL use.
    2. We study consumer responses to new Fintech credit products, adding to the existing literature studying consumer spending responses to income and liquidity.
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