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

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Harvard Business School

## Background and Motivation

### Consumers have substantial demand for short term unsecured credit



Source: Federal Reserve Bank of New York

As is evidenced by the close to \$1 Tr in outstanding unsecured consumer loans in the U.S.

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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
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- 25% of the purchase price is required at the time of purchase.
- 3 additional equal payments are required every 2 weeks thereafter.
- i.e. suppose I purchase a \$100 sweater using BNPL "pay-in-4"
  - $\cdot$  \$25 is paid at the date of purchase
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- On the one hand, in a setting with credit constraints, additional access to credit can help smooth consumption.
- On the other hand, additional access to credit can cause overspending relative to long run preferences if consumers have:
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  - An incomplete understanding of contract terms
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Background
## In the U.S. there are 5 main BNPL providers that capture >95% of of the market

BNPL Provider	Credit Check	Pay-in-4 Option?	Late fees	Non Pay-in-4 Option?	Interest Rates	Negative Reporting	Positive Reporting
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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Integration with top 10k Websites



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Data & Patterns of BNPL Use

- 1. We are able to observe BNPL use through bank account and credit card transactions level data for around 10 million active users in the U.S.
- 2. We identify BNPL transactions by making use of merchant classification provided by the data aggregator plus manual searches of transactions descriptions.
- 3. We also select a random sample of 200k BNPL and 200k non-BNPL users within the same city of residence and income class to conduct our main analysis.

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## Population summary statistics by BNPL use/year

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		As of December						
	20	17	203	18	201	9	2020	
	BNPL	user	BNPL	BNPL user BN		BNPL user		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, %							
Renter	14	14.6	14.1	16.8	15.1		
Credit card user	40.1	28.3	32.2	25.6	28.7		
Active saver	11.8	12.8	11.2	10.7	10.8		
\$400 buffer	71.8	60.4	62.8	57.9	60.9		
Paid overdrafts	7.9	11.7	10.6	12.8	11.1		
Median, \$							
Salary	4,199	3,311	3,524	3,211	3,421		
Essential spending	908	836	859	815	907		
Discretionary spending	602	574	594	565	573		
Bills	475	428	439	433	449		
Retail spending	562	547	547	555	644		
Credit transactions	7,447	5,817	6,193	5,670	5,947		
Debit transactions	7,929	6,241	6,593	6,076	6,326		
Users (out of 260,035)	16,745	14,570	10,577	2,242	4,628		







- Spend more on non-essential goods
- Shop more online
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BNPL adoption by provider: All

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## Most people use BNPL only a few times



#### Most people use BNPL only a few times



### People use BNPL more during the holiday period



What should we expect?

- 1. Consumers will optimally increase current consumption via intermporal substitution effects and a reduction in precautionary savings motive
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We investigate both effects in detail.

BNPL access and spending

- $\cdot$  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:

$$y_{it} = \alpha_{it} + \sum_{k=-12}^{24} \gamma_k \mathbb{1}\{\text{First}\_\text{BNPL}\_i - t = k\} \times Tit + \varepsilon_{it}$$

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Panel A: Retail spending



Panel B: Essential spending



Panel C: Discretionary spending



Panel D: Other spending















## 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.

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



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# We use this binary exposure instrument to estimate coefficients in the structural equation:

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

i.e. we use *E<sub>i,t</sub>*, a binary exposure variable, to instrument for Post<sub>it</sub>, which is an indicator variable equal to one after the first time a person uses BNPL

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i.e. we use  $E_{i,t}$ , a binary exposure variable, to instrument for  $Post_{it}$ , 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

Retailer	
Target	
Bed Bath & Beyond	
Michaels	
	22.4
	4.5
Nordstrom	

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

Retailer	Customers %	Revenue 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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#### Post-BNPL availability, total spending increases

Total Spend						
	Fixed Effects Reduced Form TSLS					
Post	40.16*** (1.39)		60.47*** (10.74)			
Exposure		7.512*** (1.479)				
KP Wald F Stat			1,163			

Retail Spend						
Fixed Effects Reduced Form TSL						
Post	20.16*** (0.37)		53.58*** (2.68)			
Exposure		6.758*** (0.449)				
KP Wald F Stat			1,220			

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

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		Total Spend				
	Credit Card User		Not a Credit Card User			
	Fixed Effects	Reduced Form	TSLS	Fixed Effects	Reduced Form	TSLS
Post	31.81*** (1.512)		—1.844 (16.68)	41.52*** (1.567)		68.26*** (10.57)
Exposure		—0.168 (1.517)			8.869*** (1.533)	
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		r	
	Fixed Effects	Reduced Form	TSLS	Fixed Effects	Reduced Form	TSLS
Post	18.75*** (0.391)		42.26*** (4.132)	20.45*** (0.394)		57.08*** (2.650)
Exposure		3.943*** (0.443)			7.468*** (0.455)	
KP Wald F Stat			734			991

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

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Retail Spend/Total				
	Fixed Effects	Reduced Form	TSLS	
Post	0.0335*** (0.0005)		0.0629*** (0.0035)	
Exposure		0.00697*** (0.00045)		
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		er	
	Fixed Effects	Reduced Form	TSLS	Fixed Effects	Reduced Form	TSLS
Post	0.0290***		0.0447***	0.0369***		0.0720***
Exposure	(0.000333)	0.00381*** (0.000517)	(0.00070)	(0.000303)	0.00806*** (0.000531)	(0.00430)
KP Wald F Stat			680.1			892.7

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

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Balance Estimate					
Fixed Effects Reduced Form TSLS					
Post	—659.4*** (63.9)		—997.2** (408.7)		
Exposure		—123.9** (51.4)			
KP Wald F Stat			1,163		

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

Overdraft Fee				
	Fixed Effects	Reduced Form	TSLS	
Post	0.00263*** (0.00026)	c.	0.00473*** (0.00161)	
Exposure		0.000588*** (0.000203)		
KP Wald F Stat			1,163	

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	Balance Estimate					
	Credit Card User			Not a Credit Card User		
	Fixed Effects	Reduced Form	TSLS	Fixed Effects	Reduced Form	TSLS
Post	-833.1*** (110.4)		—894.2 (1.136)	-462.9*** (54.49)		—1.663*** (360.9)
Exposure		—81.46 (103.5)			—216.0*** (47.53)	
KP Wald F Stat			690.4			959.3

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

	Overdraft Fee					
	Credit Card User			Not a Credit Card User		
	Fixed Effects	Reduced Form	TSLS	Fixed Effects	Reduced Form	TSLS
Post	0.00157*** (0.000242)		-0.00123 (0.00227)	0.00269*** (0.000361)		0.00458** (0.00232)
Exposure		-0.000112 (0.000207)			0.000595* (0.000304)	
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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We find that BNPL access is associated with:

- 1. Increased spending
- 2. A flatter relationship 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"

BNPL disproportionately provides liquidity for retail purchases

 $\rightarrow$ 

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



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the additional liquidity "sticks" in retail and leads to higher retail consumption

Ad-hoc budgeting rules or mental accounting combined with consumer myopia i.e.

- 1. 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.

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- smoother spending relative to income
- increased allocation towards retail goods
- 2. 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"
  - · Increased retail liquidity leads to increased retail spending
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