## Is There an Nth of the Month Effect? The Timing of SNAP Issuance, Food Expenditures, and Grocery Prices

### Jacob Goldin Tatiana Homonoff Katherine Meckel

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- Many public assistance benefits are delivered to recipients as a monthly lump sum
  - Benefits can be issued on one date or staggered across households throughout the month
- This paper looks at the effect of staggering benefit issuance for the Supplemental Nutritional Assistance Program (SNAP)
- We use high-frequency household and retail scanner data to estimate the effect of staggering SNAP issuance date on:
  - Food purchasing patterns
  - Price response of retailers

- Supplemental Nutrition Assistance Program (SNAP)
  - Provides food vouchers to low-income households
  - 47 million participants, total spending of \$75B per year
- Benefit issuance
  - Each household receives benefits via an electronic benefit transfer (EBT) card as a monthly lump sum
- Recipients can choose to consume these benefits all at once or steadily throughout the month

## First of the Month Effect

- NPR: Low-income families stock up at Walmart on the first of the month when they have food stamps in hand
  - Referred to as the "First of the Month Effect"



 Researchers have also documented evidence of this effect Castner & Henke (2011), Shapiro (2006), Hastings & Washington (2010) • As a result, policymakers have suggested staggering SNAP benefit issuance across households throughout the month

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  - Reduce SNAP-induced price response
    - If food prices are higher on the first of the month ⇒ incidence of SNAP benefits go to the *retailer*, not the recipient

- Food APS Dataset
  - Roughly 5,000 households interviewed between April 2012 and January 2013
  - 7-day household food expenditure diary
- Rich demographic information for individuals and households
  - Self-reported and adminstrative SNAP status

  - State of residence
    Detailed food environment information for each household

Issuance Policy	State Abbreviation
First Day Only	ID, ND, NV, OK, RI, VA, VT
Week 1 Only	CT, MT, NE, NH, NJ, WY
Week 2 Only	ME, SD
Weeks 1 and 2	AR, AZ, CA, CO, DC, DE, FL, GA, IA, IN, KS, KY, LA, MA, MD, MN, NY, OH, OK, OR PA, SC, TN, TX, UT, WA, WI, WV

Spread (3+ Weeks) AL, IL, MI, MO, MS, NC, NM

Source: USDA Food and Nutrition Services

	Full Sample	SNAP HHs	Poor Non-SNAP HHs	Non-Poor HHs
Number of Households	4,826	1,581	1,195	2,048
Number of Individuals	14,317	5,414	3,335	5,564
Household Demographics				
Female (%)	67.6	73.5	66.2	66.8
Age	49.8	45.9	52.9	49.7
HS Dip∣oma (%)	90.3	74.0	81.5	95.9
Non-white (%)	23.8	40.9	28.2	19.3
Number of Children	0.68	1.03	0.61	0.63
Married (%)	44.2	22.0	29.8	52.3
Hispanic (%)	12.7	24	18.5	9.0
Household size	2.42	2.90	2.17	2.39
Daily Food Expenditure (\$)	17.29	13.80	11.29	19.51

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### Model

$$y_{iw} = \sum_{w=2}^{4} [\alpha_w week_w + \beta_w SNAP_i * week_w] + \gamma SNAP_i + \theta_i$$

- Obs-level: household *i* in calendar week *w*
- $y_{iw}$  = average daily household food expenditure (\$/log)
- $week_w = 1$  if in the  $w^{th}$  week of the month
- SNAP = 1 if SNAP recipient
- Controls: HH/respondent demographics (HH size, children, race, education, age)

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- SNAP = 1 if SNAP recipient
- Controls: HH/respondent demographics (HH size, children, race, education, age)
- Restrict sample to households below 185% FPL
  - $\alpha_w$ : change in daily food expenditure between the first and the  $w^{th}$  week of the calendar month for poor, non-SNAP households
  - $\beta_{\rm w}:$  additional change in daily food expenditure for SNAP recipients relative to poor non-recipients

## Predictions

• Key assumption: interview date randomly assigned



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- If intra-month cyclicality in food expenditure is due to the SNAP program, then we would predict:
  - First-of-the-month states:  $0 > \beta_2 > \beta_3 > \beta_4$
  - Staggered issuance states:  $0 = \beta_2 = \beta_3 = \beta_4$

	Food Expenditure (\$)			
Issue Policy:	1st Week	Week 1-2	3+ Weeks	
SNAP*Week 2 ( $\beta_2$ )	-13.10**	0.548	-2.400	
	(6.151)	(2.371)	(4.386)	
SNAP*Week 3 ( $\beta_3$ )	-16.50**	-3.092**	-2.594	
	(6.742)	(1.933)	(4.055)	
SNAP*Week 4 ( $\beta_4$ )	-16.68***	7.878***	-2.265	
	(6.042)	(1.902)	(5.668)	
Week 2 ( $\alpha_2$ )	-3.195	1.995	-1.038	
	(2.416)	(1.352)	(2.254)	
Week 3 ( $lpha_3$ )	3.524	0.498	-1.773	
	(2.828)	(1.204)	(1.930)	
Week 4 ( $lpha_4$ )	0.106	1.429	0.199	
	(2.895)	(1.280)	(2.509)	
SNAP	12.52**	4.556***	5.031	
	(5.050)	(1.476)	(3.542)	
DV Mean	13.53	14.79	13.30	
N	373	3,278	855	

\* ho < 0.10 , \*\* ho < 0.05 , \*\*\* ho < 0.01

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Food Expenditure (\$)			Log Expenditure			
Issue Policy:	1st Week	Week 1-2	3+ Weeks	1st Week	Week 1-2	3+ Weeks
SNAP*Week 2 ( $\beta_2$ )	-13.10**	0.548	-2.400	-0.622	0.0176	-0.074
	(6.151)	(2.371)	(4.386)	(0.474)	(0.172)	(0.272)
SNAP*Week 3 ( $\beta_3$ )	-16.50**	-3.092**	-2.594	-1.283***	-0.231	-0.022
	(6.742)	(1.933)	(4.055)	(0.446)	(0.175)	(0.265)
SNAP*Week 4 ( $\beta_4$ )	-16.68***	-7.878***	-2.265	-1.194***	-0.528***	-0.174
	(6.042)	(1.902)	(5.668)	(0.402)	(0.187)	(0.368)
Week 2 $(\alpha_2)$	-3.195	1.995	-1.038	-0.253	0.105	-0.109
	(2.416)	(1.352)	(2.254)	(0.294)	(0.114)	(0.177)
Week 3 $(\alpha_3)$	3.524	0.498	-1.773	0.528**	-0.0507	-0.275*
	(2.828)	(1.204)	(1.930)	(0.239)	(0.124)	(0.165)
Week 4 ( $lpha_{4}$ )	0.106	1.429	0.199	0.236	0.120	-0.0804
	(2.895)	(1.280)	(2.509)	(0.260)	(0.123)	(0.236)
SNAP	12.52**	4.556***	5.031	0.620**	0.276**	0.280
	(5.050)	(1.476)	(3.542)	(0.290)	(0.131)	(0.210)
DV Mean	13.53	14.79	13.30	2.23	2.40	2.32
N	373	3,278	855	299	2,615	652

\*  $\rho < 0.10$  , \*\*  $\rho < 0.05$  , \*\*\*  $\rho < 0.01$ 

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- Placebo test:
  - Repeat analysis restricting sample to non-SNAP households
  - Compare food expenditure patterns of poor (<185% FPL) vs. non-poor households

$$y_{iw} = \sum_{w=2}^{4} [\alpha_w week_w + \beta_w Poor_i * week_w] + \gamma Poor_i + \theta_i$$

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### Placebo: Poor vs. Non-Poor Non-Recipients

		E×penditure (\$	)
SNAP Policy:	1st Week	Week 1-2	3+ Weeks
Poor*Week 2	-3.155	-0.795	2.935
	(3.974)	(2.142)	(3.204)
Poor*Week 3	7.172	1.050	-5.487
	(4.402)	(1.972)	(3.355)
Poor*Week 4	2.104	0.187	-1.199
	(4.075)	(2.144)	(4.337)
Week 2	2.753	2.571	-4.038*
	(3.097)	(1.582)	(2.154)
Week 3	-0.713	- 1.026 2.	
	(2.696)	(1.505)	(2.696)
Week 4	-0.420	0.965	0.212
	(2.923)	(1.652)	(3.347)
Poor	-5.396**	-4.332***	-6.151**
	(2.673)	(1.455)	(2.476)
DV Mean	15.49	17.78	15.21
N	557	3,834	906

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• Household Expenditure Data: Nielsen Consumer Panel Data

- Transaction-level expenditure data from 2004-2011
- 60,000 households per year
- Average time in panel: 4 years
- Annual demographic information to estimate SNAP eligibility



$$y_{iwmy} = \sum_{w=2}^{4} [\alpha_w week_w + \beta_w SNAP_{iy} * week_w] + \gamma SNAP_{iy} + \delta_m + \eta_y + \theta_i$$

- Obs-level: household *i* in week *wmy*
- y<sub>iwmy</sub> = log(expenditure)
- $week_w = 1$  if in the  $w^{th}$  week of the month
- *SNAP* = 1 if SNAP-*eligible*
- Fixed effects: month  $(\delta_m)$ , year  $(\eta_y)$ , and household  $(\theta_i)$

## Food Expenditure Patterns by Issuance Policy

#### Food Expenditure



## Food Expenditure Patterns by Issuance Policy



## Food Expenditure Patterns by Issuance Policy





- So far relied on cross-sectional variation in policy across states
- Case study: Illinois
  - Before February 2010: (most) benefits issued on the 1st
  - March 2010 to June 2013: issued from the 1st to 23rd
  - July 2013 to February 2014: issued from the 1st to 10th
  - March 2014 to present: issued from the 1st to 23rd
- Administrative data from Illinois, 2008-2014
  - Daily benefit issuance and redemption

## Illinois: Issuance and Redemption Patterns



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- Is there SNAP-induced cyclicality in *total* food sales? If so...
  - Question: Do retailers respond by changing prices in high demand weeks?

- Is there SNAP-induced cyclicality in *total* food sales? If so...
  - Question: Do retailers respond by changing prices in high demand weeks?
- Possible pricing models
  - Cyclical pricing: higher prices during peak demand Hastings & Washington (2010)
  - Counter-cyclical pricing: lower prices during peak demand (e.g., loss-leader) chevalier, Kashyap, & Rossi (2003)
  - No price response: reputation concerns, price sensitivity of SNAP recipients Kaplan & Menzio (2013)

	Full Sample	SNAP	Poor, Non-SNAP	Non-Poor
Reason for shopping at primary store (%)				
Low Prices	52.7	60.3	56.7	50.1
Close to Home	52.8	46.9	44.3	56.1
Variety	25.4	19.5	22.6	27.2
Quality	21.0	18.0	16.2	22.9

Source: Food APS

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### Price Data: Nielsen Retailer Data

#### Nielsen Retail Scanner Data

- Weekly store-level price and volume data by UPC
- Years: 2006-2012
- Store characteristics: Retail chain (e.g., Safeway), zip code

	Grocery Stores
Number of Stores	10,070
Average Annual Food Sales (\$)	7,941,396
Total Sales Volume (%)	53
Number of retailers	90
% SNAP in Zip Code	10.1

Source: Nielsen Retailer Scanner data, 2006-2012

• Create a share-weighted price index for each store-day:

$$log(P_{st}) = \sum_{k} \omega_k log(p_{kst})$$

- $p_{kst} = price \text{ for product (UPC) } k \text{ on date } t \text{ in store } s$
- $\omega_k = \text{total share spent on product } k$  among SNAP-eligible households in the Consumer Panel

(B)

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$$Y_{st} = \beta \, FracIssue_{swmy} + \mu_w + \delta_m + \eta_y + \theta_s$$

- $Y_{st} = \log(\text{food sales}) \text{ or } \log(\text{price})$
- FracIssue = fraction of benefits issued in store s's state in week w
- Fixed effects: week  $(\mu_w)$ , month  $(\delta_m)$ , year  $(\eta_y)$ , and store  $(\theta_s)$
- Interpretation: If 100% of a state's SNAP benefits are issued in a given week, food sales/prices are  $\beta$ % higher in that week

- If 100% of a state's SNAP benefits are issued in a given week...
  - Total food sales are **5.8%** higher in that week (vs. weeks in which no snap benefits are issued)
  - Food prices are 0.07% higher in that week

	log(Food Sales)	log(Price)
	(1)	(2)
FracIssue	0.0580***	0.0007*
	(0.0107)	(0.0004)
N	22,850,351	22,850,351

\* ho < 0.10 , \*\* ho < 0.05 , \*\*\* ho < 0.01

Source: Nielsen Retail Scanner Data, 2006-2012.

## Retailer Response by Local SNAP Prevalence

Jacob Goldin

If 100% of SNAP benefits issued in a week, how do food sales and prices differ in that week?



## Retailer Response by Local SNAP Prevalence

Jacob Goldin

If 100% of SNAP benefits issued in a week, how do food sales and prices differ in that week?



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	Full Sample	SNAP	Poor, Non-SNAP	Non-Poor
Drives to Store (%)	87.8	65.8	78.0	94.7
Time to Store (min)	10.9	11.5	12.4	10.3
Distance to Store (miles)	5.1	4.4	5.0	5.3
Number of Stores	3.4	3.3	3.2	3.5

Source: Food APS

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# Retailer Response by Market Concentration



• Slightly larger, though still economically small price response in stores in high-SNAP zip codes with few local competitors

- Expenditure: SNAP issuance policy significantly affects cyclicality in aggregate food demand
  - Aggregate food expenditure of SNAP households peaks at the start of the month in states that do not stagger SNAP issuance
  - No aggregate cyclicality in states that stagger issuance

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  - Aggregate food expenditure of SNAP households peaks at the start of the month in states that do not stagger SNAP issuance
  - No aggregate cyclicality in states that stagger issuance
- Prices: no evidence that retailers respond to these predictable increases by changing food prices
- Policy implications of SNAP issuance staggering:
  - Likely to reduce surges in customer traffic
  - Unlikely to substantially affect food prices/incidence of benefits

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- Questions? tah297@nyu.edu