

#### Personalized Mobile Targeting with User Engagement Stages: Combining Structural Forward-Looking Hidden Markov Model and Field Experiment



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#### Apps are trying to understand users





Hello and Welcome to GrubHub! Fill out your profile for a 20% off coupon.

#### Best Buy 16m ago

View today's Deal of the Day. Quantities are limited.

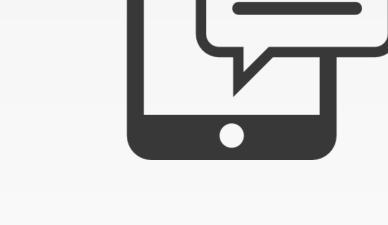
#### Facebook 3m ago

Your last post was 15 days ago. Tell friends what you're up to now.

#### Level 5m ago

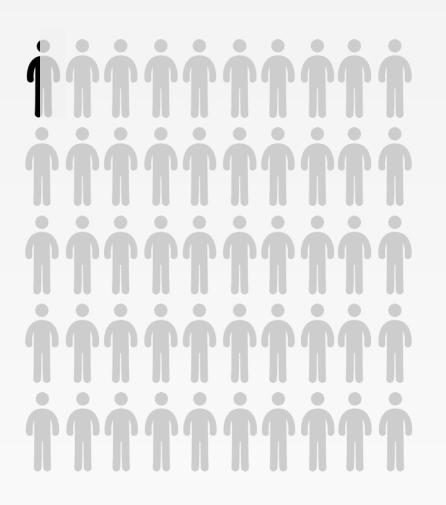
Yingjie! You've spent less than 50% of your Spendable this week! Good job!







## Are their strategies effective?



In 2015, the average mobile app retention rate was 42% after the first month, and 25% after the first three months.

2% of users paid for in-app content

Half of the revenues were contributed by 0.2% of users (Feb 2016)



Can we design better marketing strategies to improve mobile user engagement?



#### What have been done in the literature?

#### User engagement

Measure with recent activities: Claussen et al. (2013), Qi et al. (2011)

Survey-based data: Kim et al. (2013)

#### Mobile app platform

Aggregated-level analysis: Garg and Telang (2013), Liu et al. (2013)

Individual-panel data: Ghose and Han (2014)



## 3-Step Research Design

- Step 1: Randomized field experiment
  - → average causal effects of different promotions
- Step 2: FHMM
  - → detection of user engagement using tapstream data
  - → heterogeneous treatment effects
- Step 3: Simulation
  - → engagement-based targeting strategies



## Experiment Design

**Treatment 1** 

Price promotion

**Treatment 2** 

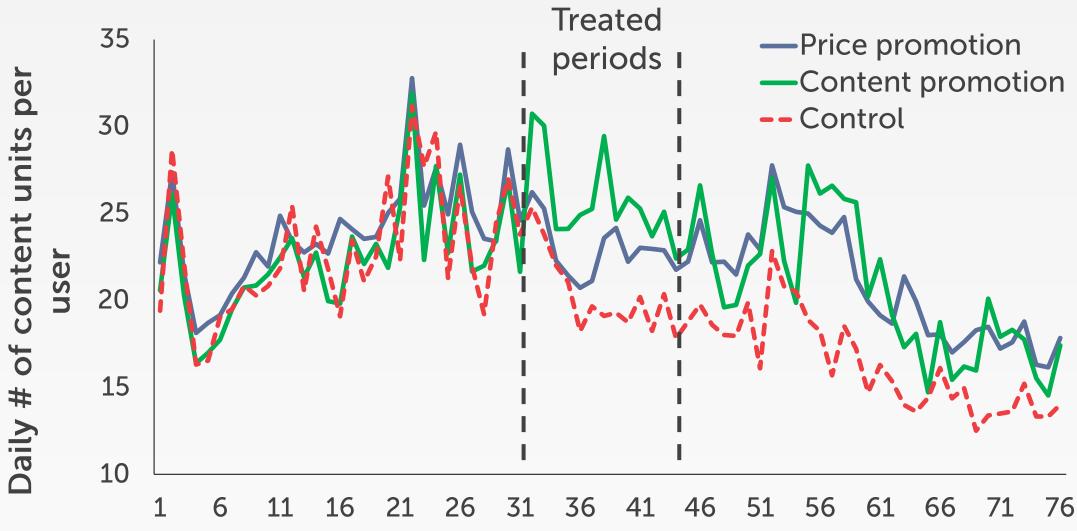
Free content promotion

Control

Non-pricing placebo messages

User ID	Time Stamp	Book ID	Book name	Genre ID	Genre Name	Payment ID	Payment	Chapter ID	<b>Chapter Name</b>	Free or not
85	7/22/2015 20:10	403696444	我的刁蛮女上	11	都市	15	按章	404509159	第七十八章落	1
20	8/3/2015 14:04	388917988	藏獒,远去的	41	纪实		免费	388918491	序言	0
504	7/201015 16:25	LISER	ID, ti	me <sup>14</sup>	stam	n co	nte	n+4115294		1
504 1087 2398	//4/2.15 9:30	386299128	红河边的花腰	26	3 COIIII	p, c	包月	386299271		1
2398	8/6/2013 4:19	infor	matic	DD 43	021/100	ont	onti	390053884		0
			macic	)   , <sub> </sub>	payii		Opti	OH		

#### Overall Treatment Effect



Time indicator



## **Average Treatment Effects**

 $Y_{it} = \alpha_0 + \alpha_1 Test_t + \alpha_2 Treat1_i \times Test_t + \alpha_3 Treat2_i \times Test_t + \alpha_t postTest_t + \alpha_5 Treat1_i \times postTest_t + \alpha_6 Treat2_i \times postTest_t + \xi_i + \varepsilon_{it}$ 

			Table 3 F	ield Experimer	nt Analysis				
$Y_{it}$	With active users only				With all users				
$_{it}$	# of	units	# of fr	ee units	# of	units	" "	ee units	
Treat1	1.0026*	0.9435*	0.1306	0.1117	1.0811***	1.0811***	0.2375***	0.2375***	
imesTest	(0.4492)	(0.4663)	(0.1374)	(0.1352)	(0.0964)	(0.0956)	(0.0328)	(0.0305)	
Treat2	0.8152*	0.7839*	0.2543*	0.213*	0.3280***	0.3280***	0.0888**	0.0888*	
imesTest	(0.4294)	(0.4453)	(0.1314)	(0.1291)	(0.1004)	(0.0995)	(0.0341)	(0.0318)	
Test	-1.1543***	-1.2993***	-0.4301***	-0.3597***	-0.3833***	-0.3833***	-0.1786***	-0.1786*	
	(0.3346)	(0.3469)	(0.1024)	(0.1006)	(0.0738)	(0.0731)	(0.0251)	(0.0234)	
Treat1		1.7882***		0.0120		1.4968***		0.2287***	
$ imes$ post ${ t Treat}$		(0.3842)		(0.1114)		(0.0703)		(0.0225)	
Treat2		1.8568***		0.0418		0.2482***		0.0012	
$ imes$ post ${ t Treat}$		(0.3670)		(0.1064)		(0.0732)		(0.0234)	
postTreat		-2.3748***		-0.3070***		-1.4124***		-0.5674***	
		(0.2890)		(0.0838)		(0.0538)		(0.0172)	
Observations	322,328	569,696	322,328	569,696	1,193,680	2,109,760	1,193,680	2,109,760	

Notes: \*\*\*p<0.001; \*\*p<0.01; \*p<0.05.

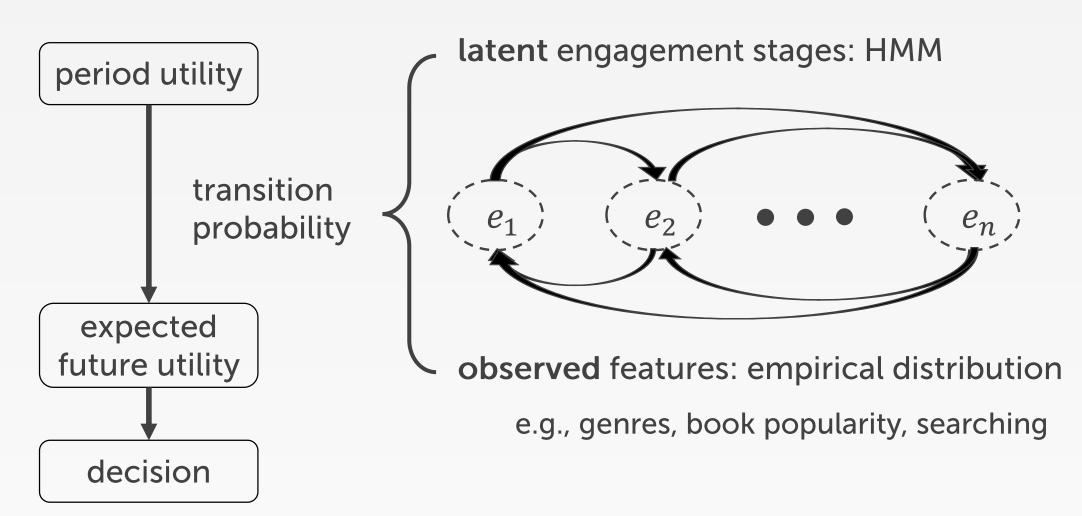


## Forward-looking Hidden Markov Model (FHMM)



Input: decision sequence, observed content features

Output: engagement stage, consumer preference





## Heterogeneous Treatment Effects

Table 7	Estimated Transition Matrix of Engagement Stages					
	$f(e' e, ar{\mathtt{CF}}, ar{\mathtt{SP}})$	e' = 1	e'=2	e' = 3	e'=4	
		(aware)	(exploring)	(active)	(addicted)	
	e = 1	0.9993	0.0002	0.0005	0.0000	
Control:	e=2	0.9771	0.0024	0.0080	0.0125	
without promotion	e = 3	0.6677	0.0071	0.2645	0.0607	
	e = 4	0.3429	0.1773	0.2580	0.2218	
	e = 1	1.0000	0.0000	0.0000	0.0000	
Treatment 1:	e=2	0.7685	0.0875	0.0040	0.1400	
price promotion	e = 3	0.2847	0.7122	0.0018	0.0013	
	e=4	0.1195	0.0565	0.2234	0.6007	
	e = 1	0.9997	0.0003	0.0000	0.0000	
Treatment 2: free-	e=2	0.5326	0.3053	0.0428	0.1194	
content promotion	e = 3	0.2901	0.0286	0.1259	0.5554	
	e=4	0.1925	0.1249	0.3819	0.2965	



## Heterogeneous Treatment Effects

Table 8 Field Experiment Analysis by Segment									
Engagement	Withou	ut post-ti	reatment	period	With post-treatment period				
Stage	e=1	e=2	e = 3	e=4	e=1	e=2	e = 3	e=4	
Treat1	1.9754*	2.8907	2.6116	6.7371*	1.9832*	2.8562	2.7984	6.2227*	
imesTest	(1.0115)	(4.1255)	(5.9846)	(3.2803)	(1.0090)	(3.9856)	(5.5721)	(3.2435)	
Treat2	1.0968*	4.0066	4.0339	7.7964*	1.2421	3.8507	4.1725	7.0252*	
imesTest	(1.0529)	(4.0760)	(5.9526)	(3.3015)	(1.0609)	(3.9359)	(5.5356)	(3.2553)	
Test	-1.4397**	4.4666	-5.0889	-7.5515**	-1.5865**	-4.2040	-5.5133	-7.3530**	
	(0.6057)	(3.9705)	(5.8600)	(3.1431)	(0.6123)	(3.8276)	(5.4367)	(3.1033)	
Treat1					2.9279*	4.4983	-0.6426	1.6779	
imespostTreat					(1.2983)	(4.9389)	(2.8639)	(2.0801)	
Treat2					3.4285**	5.56678	-0.8956	2.0980	
imespostTreat					(1.1628)	(4.9058)	(2.8279)	(2.1053)	
postTreat					-3.0413***	-4.4841	-2.3989	-1.7587	
					(0.6813)	(4.8072)	(2.6000)	(1.8171)	
Observations	158,928	56,330	58,265	54,524	280,896	99,560	102,980	96,368	



## Targeting-Strategy Design

Baseline: mass promotion

Experience-based personalized

K-means-based

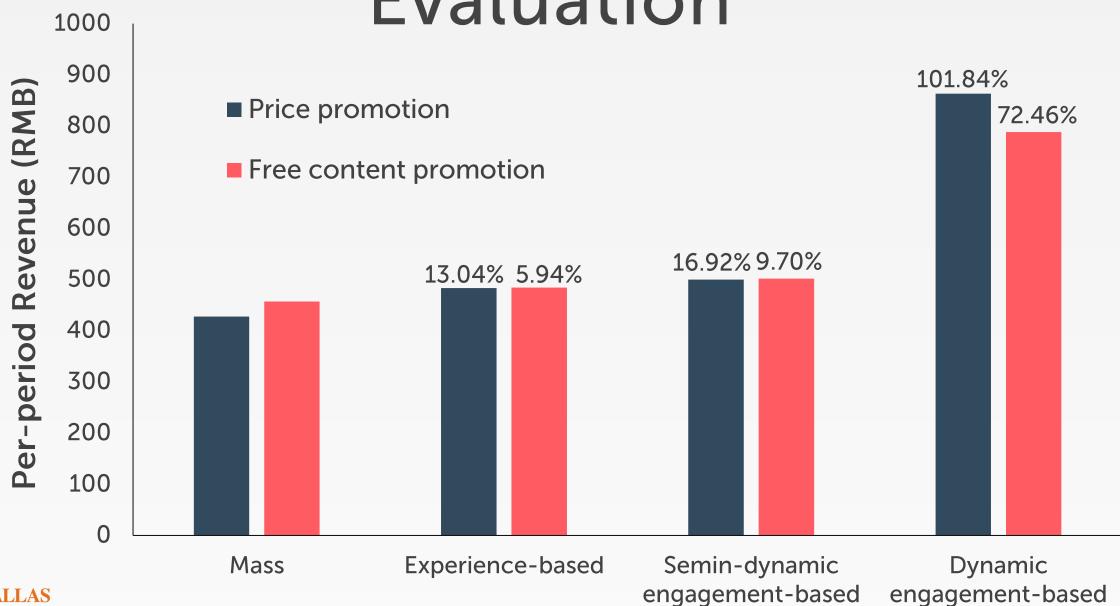
Myopic-HMM-based

Semi-dynamic engagement-based

Dynamic engagement-based



### Evaluation





#### Contributions

Methodological contributions



A structural model: FHMM

A randomized field experiment

A methodology combination

Managerial contributions



An effective approach of personalizing the interventions based on FHMM



# Thank You Q&A

