

Learning Thresholds for PV Change Detection from Operators' Labels

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Jiao Luo, Xiaowei Jing, Mei Feng



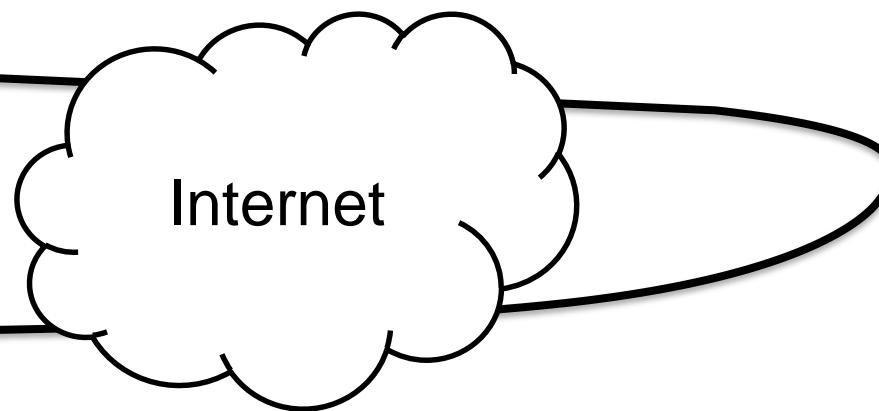
PVs (Page Views)



When you submit a query and get the result, you contribute a **search PV (Page View)**



User



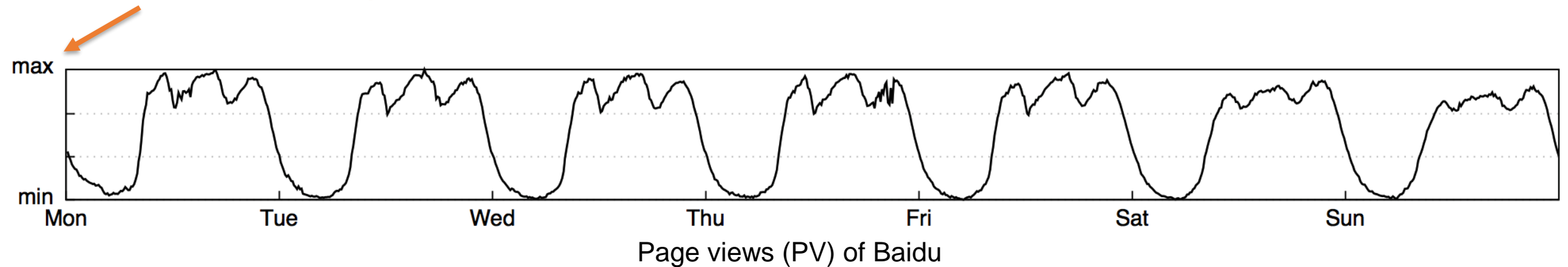
Server



PVs (Page Views)

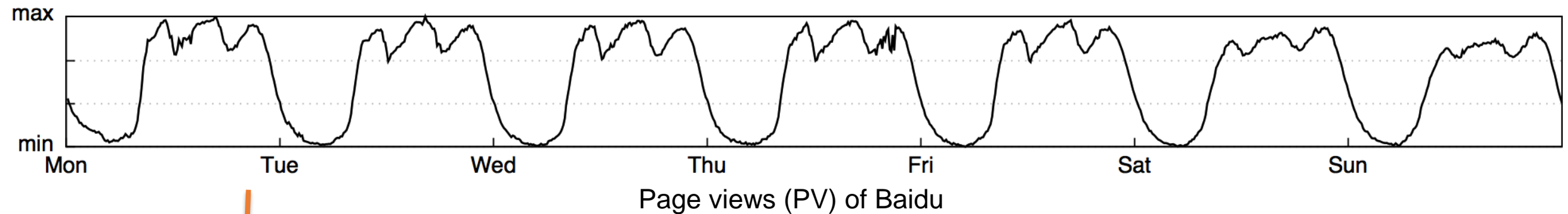
- **PV** is crucial for search engines (e.g., revenue and market share)

Over a million PVs per minute



PVs (Page Views)

- **PV** is crucial for search engines (e.g., revenue and market share)

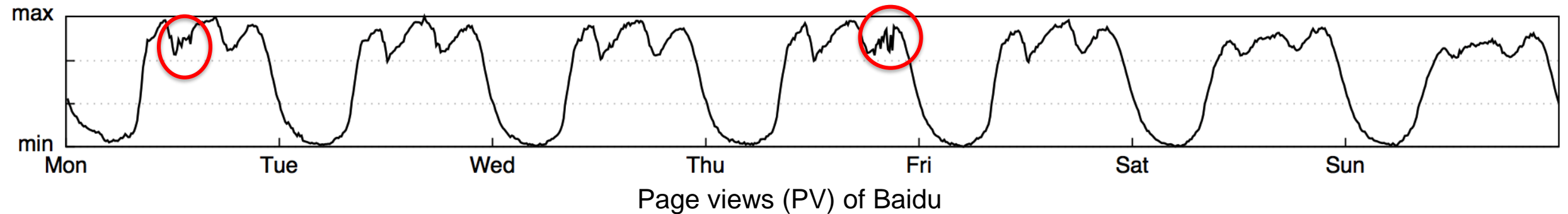


PV breakdowns
in different ways

- ISPs (e.g., Nanjing Telecom, Beijing Unicom)
- Devices (e.g., mobile phones, PCs)
- Data centers
- ...

PVs (Page Views)

- **PV** is crucial for search engines (e.g., revenue and market share)



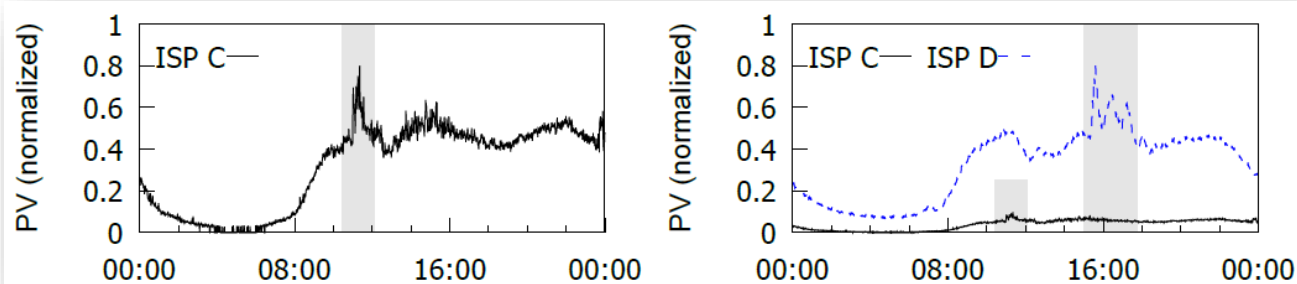
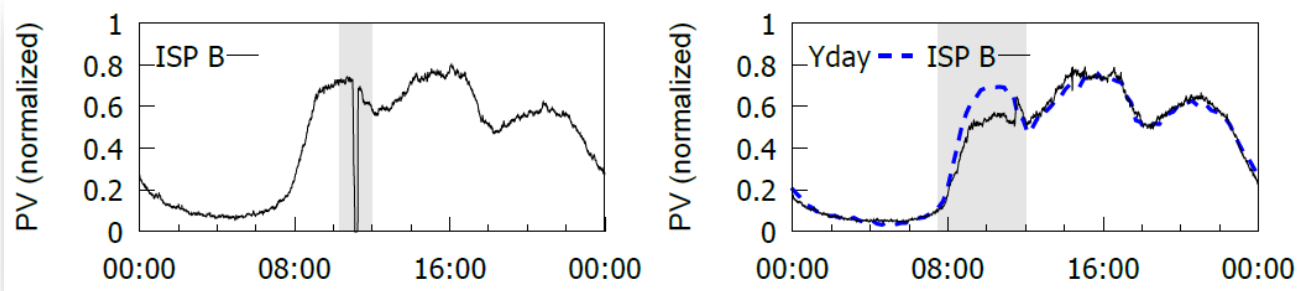
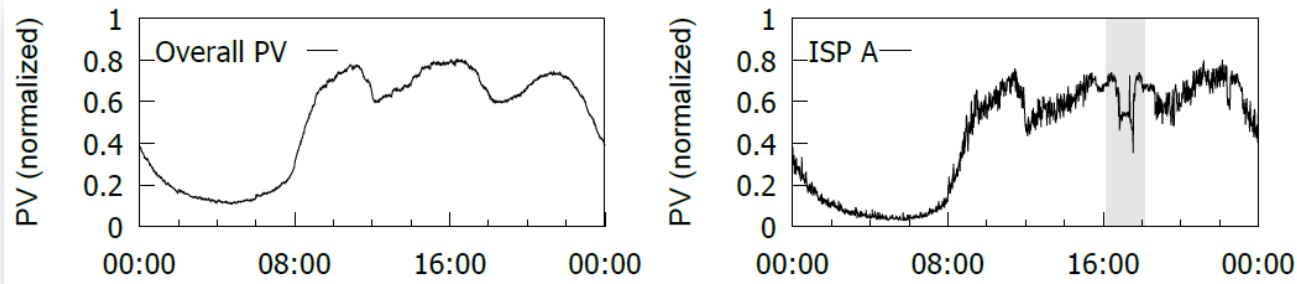
Significant PV changes → Potential failures, bugs, attacks...



Operators are responsible for detection

What changes do they care?

Significant PV Changes



Overall PV is not efficient

- Fine-grained PVs from 100+ ISPs

Change properties

- Change severity & duration

ISP PVs are not born equal

- Large ISP is more important

Problem

Manfully tune **many**
change severity & duration thresholds
for different ISPs

Overall PV is not efficient

- **Fine-grained PVs from 100+ ISPs**

Change properties

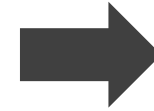
- **Change severity & duration**

ISP PVs are not born equal

- **Large ISP is more important**

Problem

Manfully tune **many**
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for different ISPs

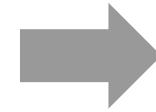


Idea

Automatically tune thresholds
from operators' feedback of
the detection results

Problem

Manfully tune **many**
change severity & duration thresholds
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Challenges

- Operators only provide **a few** labels
- Labels can be **inaccurate** sometimes

Significant PV Changes

Problem

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Idea

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Solution

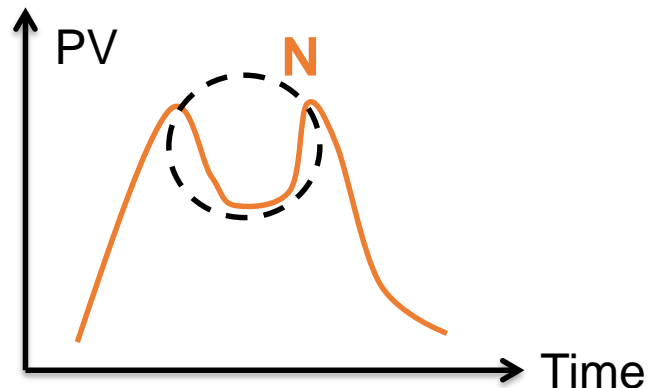
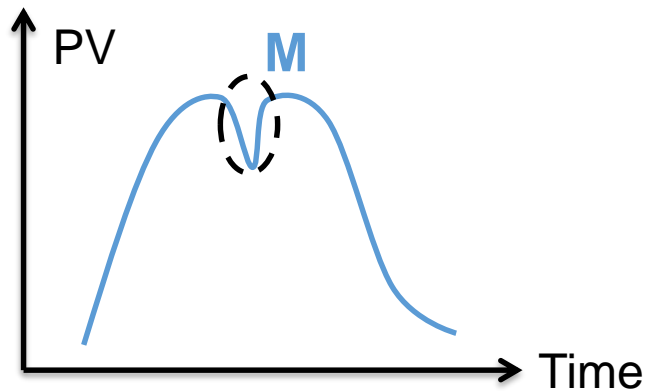
Make use of every label

Challenges

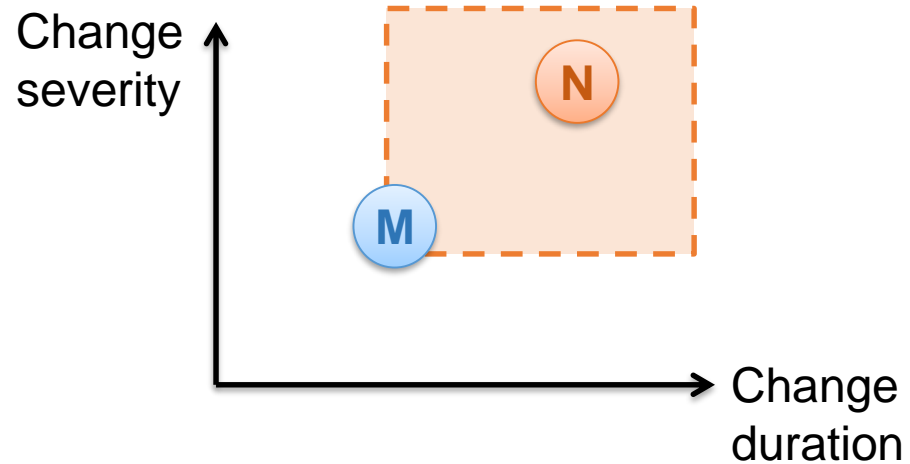
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Observation 1

- Change significance **monotonically increases** with the severity and the duration (for the same ISP)

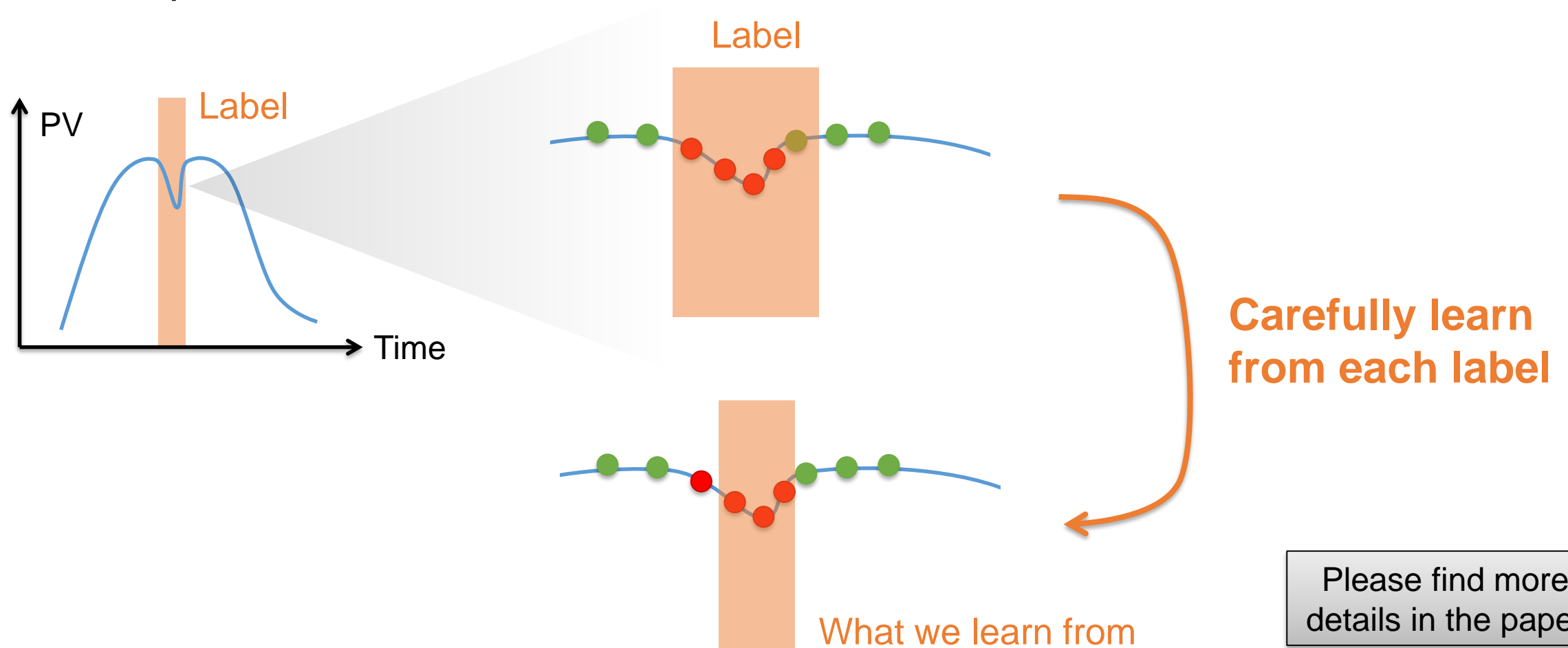


If **M** is a significant change, so should be **N**



Observation 2

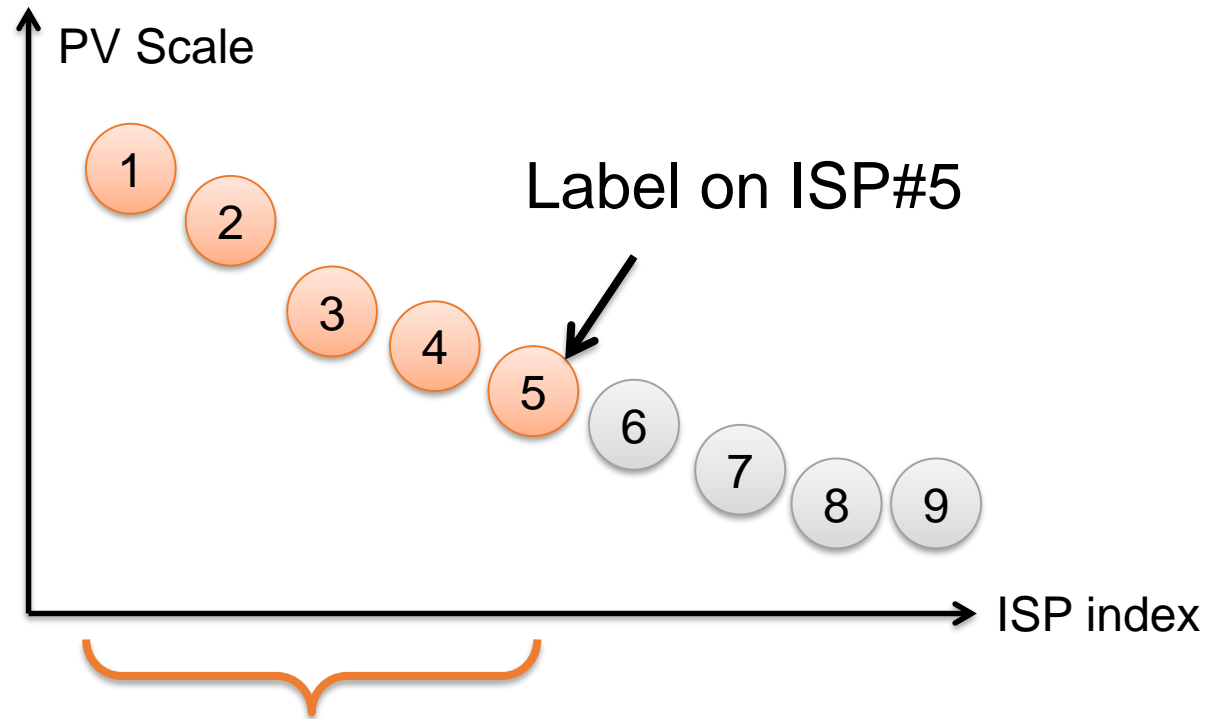
- Most of the label interval is correct, though not 100%
 - Assumption: over a half



Please find more details in the paper

Observation 3

- Larger ISPs are more important → **Sharing learning outcomes**

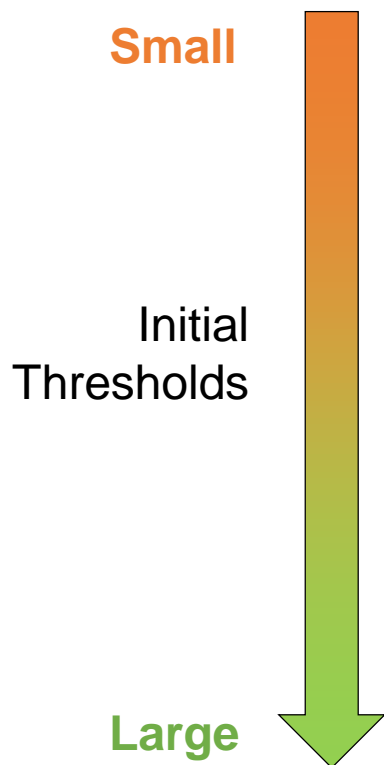


Share the learning result among larger ISPs

Please find more details in the paper

Result # 1

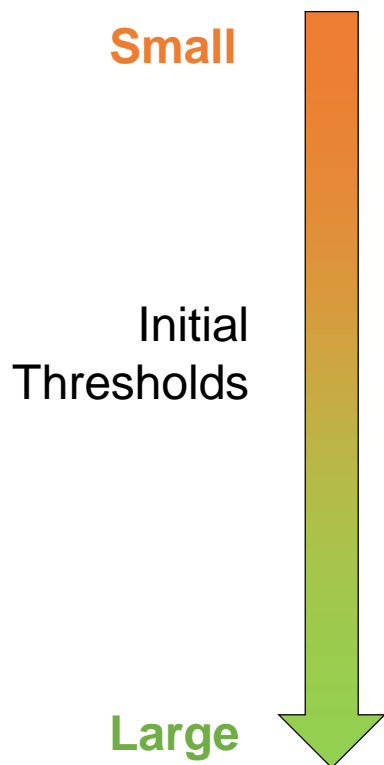
- Ground truth: real tickets
- 4-month overall PV from Baidu



Initial \mathcal{T}	Threshold Learning	#Alarms	Identified Tickets(%)	#Labels
{[2, 1.96]}	Fixed Thld	3744	100%	-
	Lazy OP	84	79%	16
	Careful OP	54	87%	25
{[3, 3.92]}	Fixed Thld	108	83%	-
	Lazy OP	39	83%	4
	Careful OP	51	83%	25
{[5, 5.88]}	Fixed Thld	18	67%	-
	Lazy OP	18	79%	4
	Careful OP	30	79%	20
{[7, 7.84]}	Fixed Thld	13	50%	-
	Lazy OP	25	58%	4
	Careful OP	24	75%	9
{[10, 9.8]}	Fixed Thld	4	17%	-
	Lazy OP	13	46%	2
	Careful OP	34	67%	19

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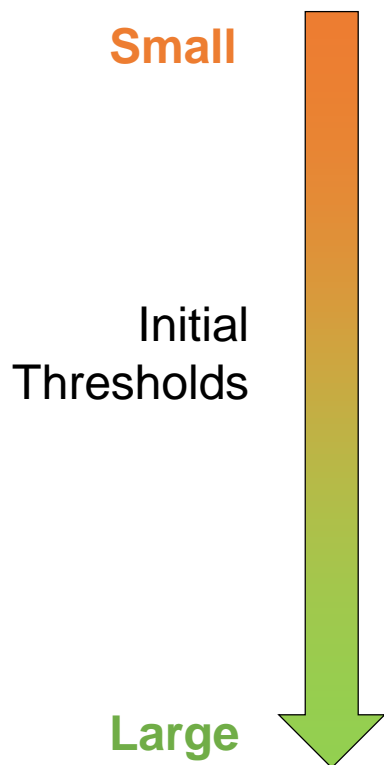
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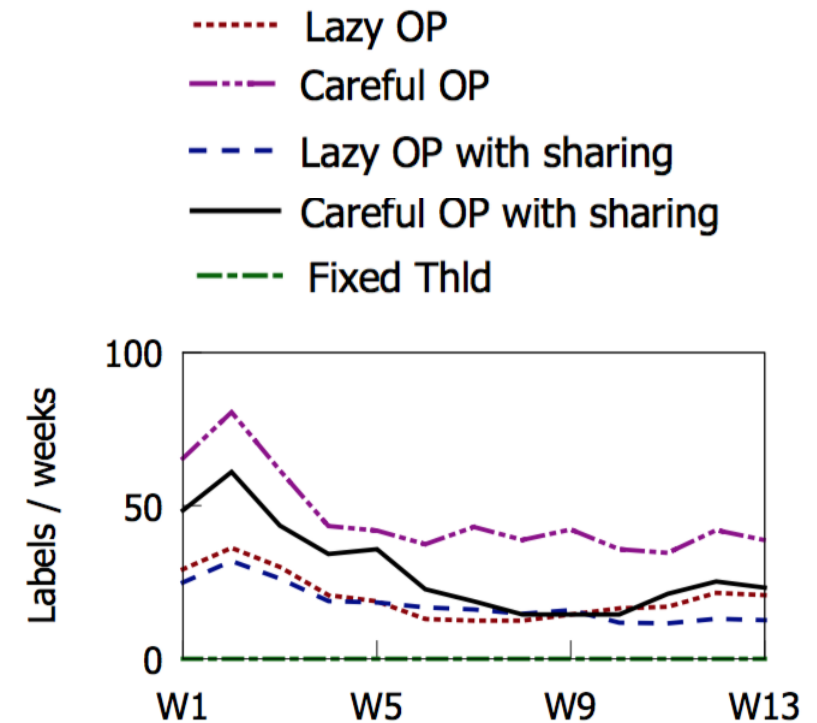
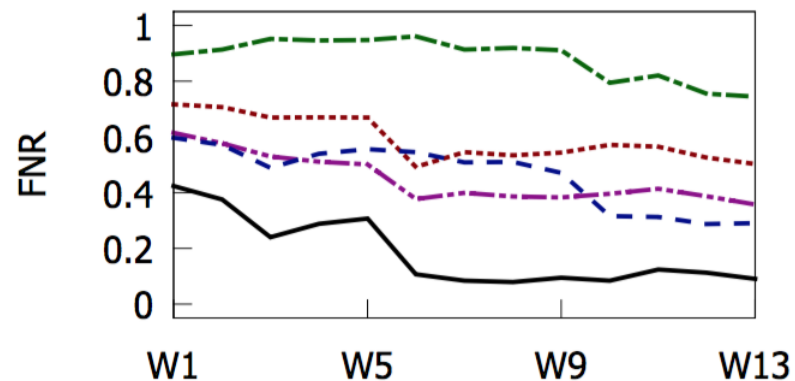
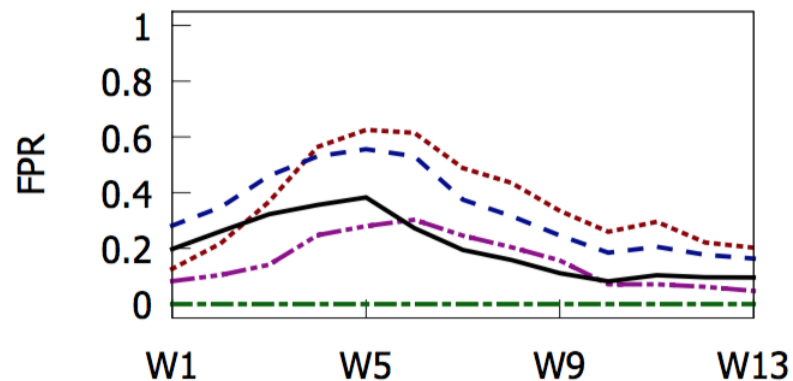
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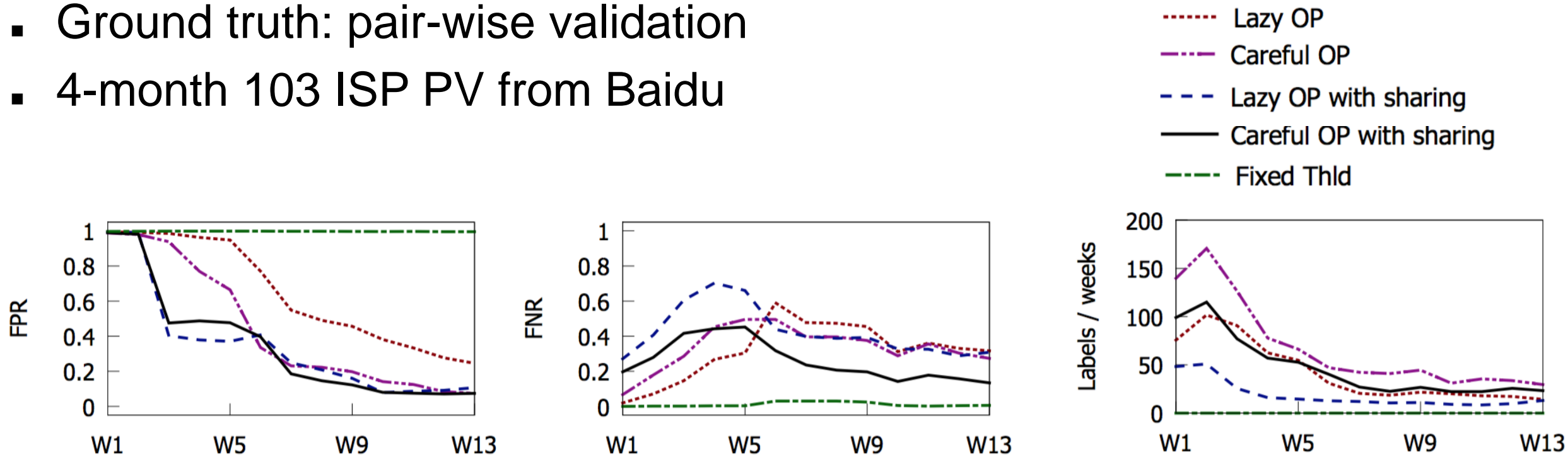
- Ground truth: pair-wise validation
- 4-month 103 ISP PV from Baidu



Large initial thresholds

Result # 2

- Ground truth: pair-wise validation
- 4-month 103 ISP PV from Baidu



Small initial thresholds

- Automatically tune hundreds of thresholds from labels
- Learn the intention behind each label as much as possible
- Increase the accuracy significantly with tens of labels a week

Thank you