

Leveraging AI for Faster Storage Access: a Graph-Neural- Network-Based Prefetcher

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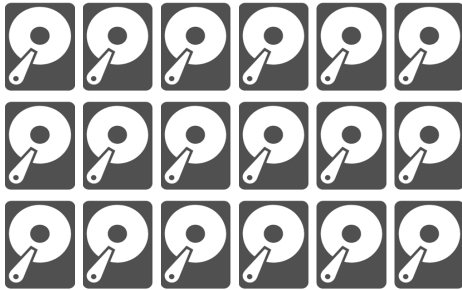



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Backend storage is slow

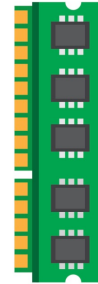
Bulk Storage


 Meta  Microsoft



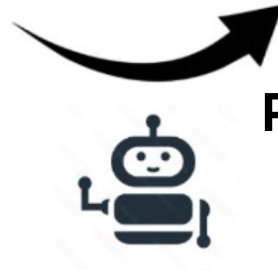

Slow
← 1.5 day

Memory




Fast
← 1 second

CPU



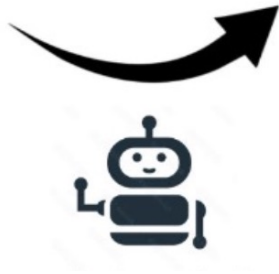
Prefetch!

Prefetching improves access speed

No prefetch



With prefetch



Previous work

['92] Stride



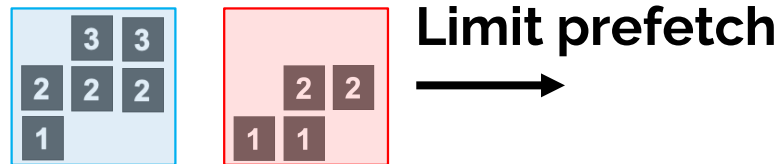
['20] Leap



['21] LSTM

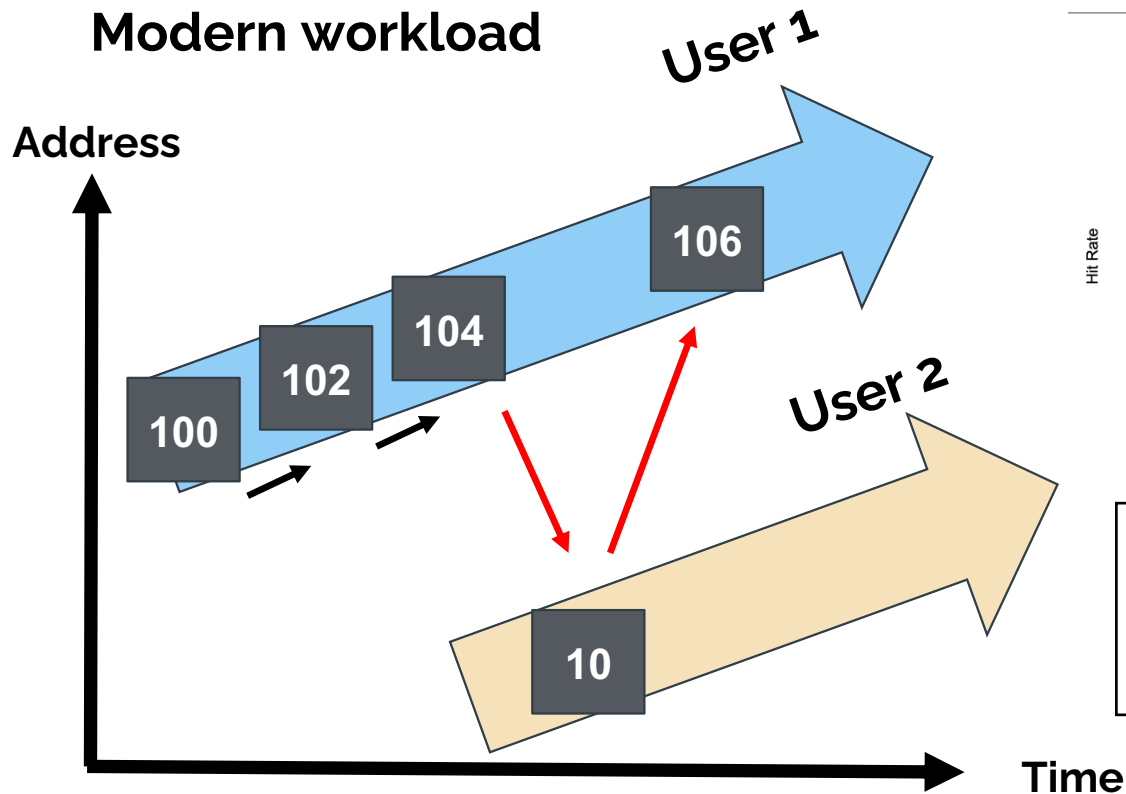


['24] Baleen



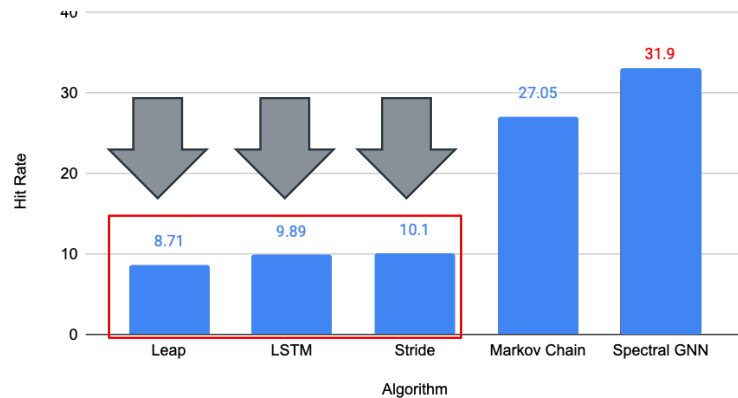
Current problem

Modern workload



Average hit rate on Alibaba, Tencent,
and Microsoft traces

Previous method



- 1) Not prone against short-term irregularities.
- 2) Can't see a global picture.

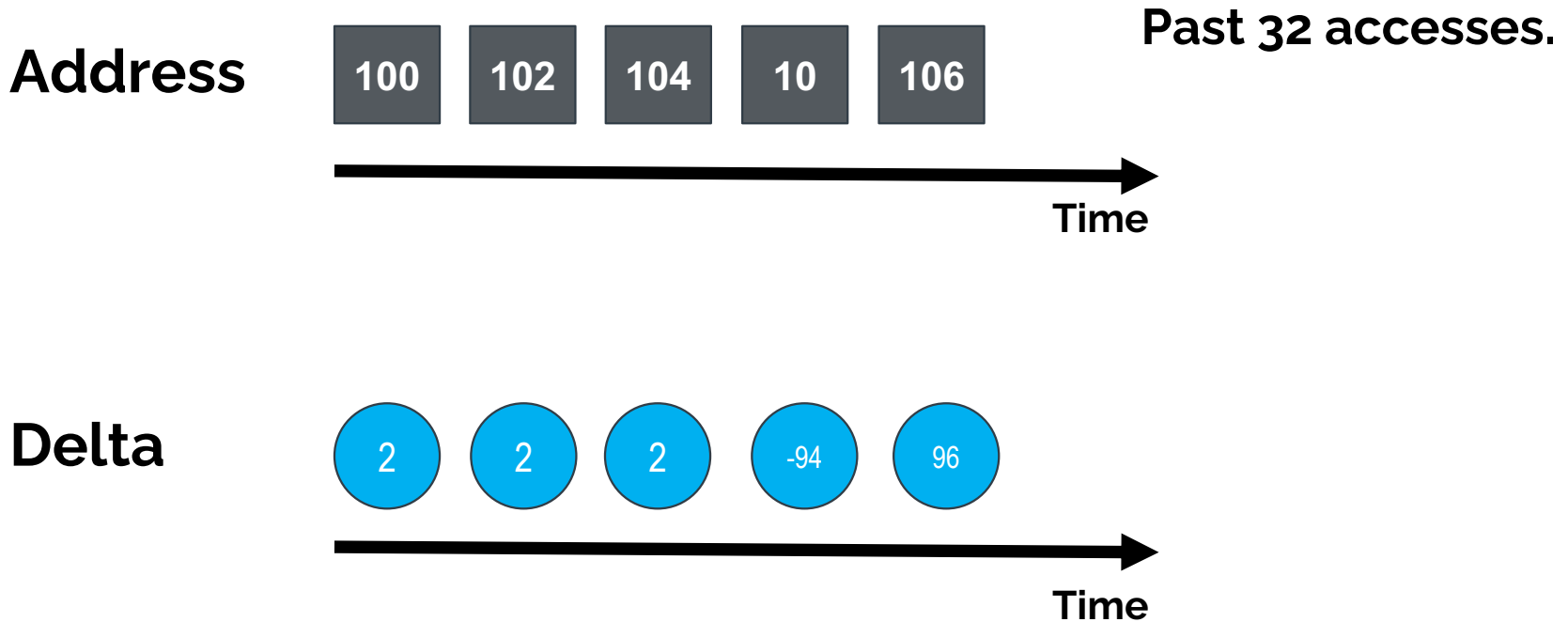


Methodology

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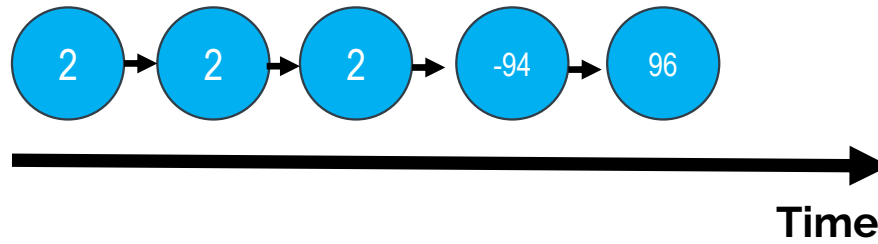


Step 1: Convert to deltas

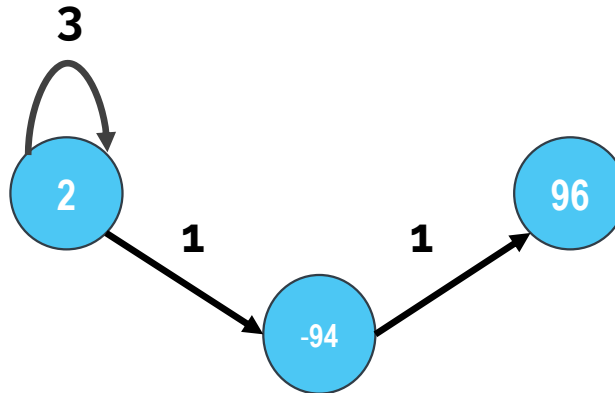


Step 2: Build a graph

Delta

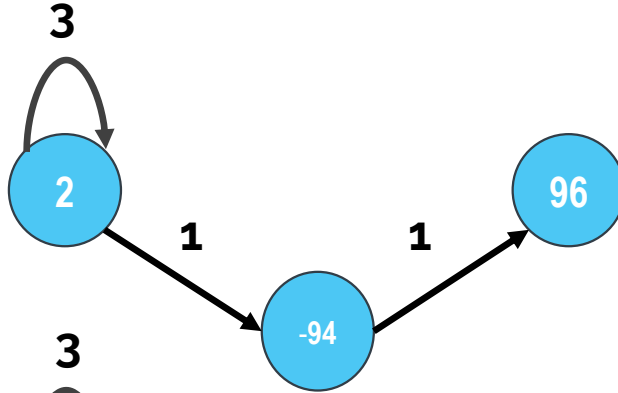


Graph



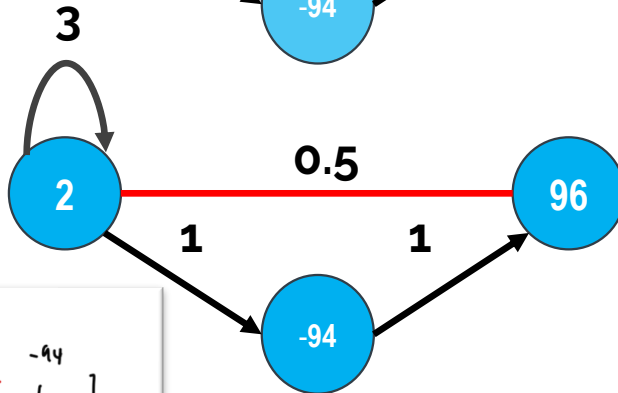
Step 3: Add full connections

Graph



Directed edges
capture temporal
information.

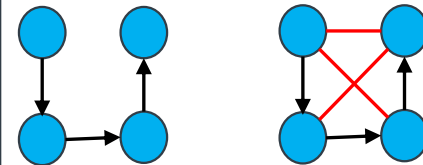
Fully-
connected
Graph



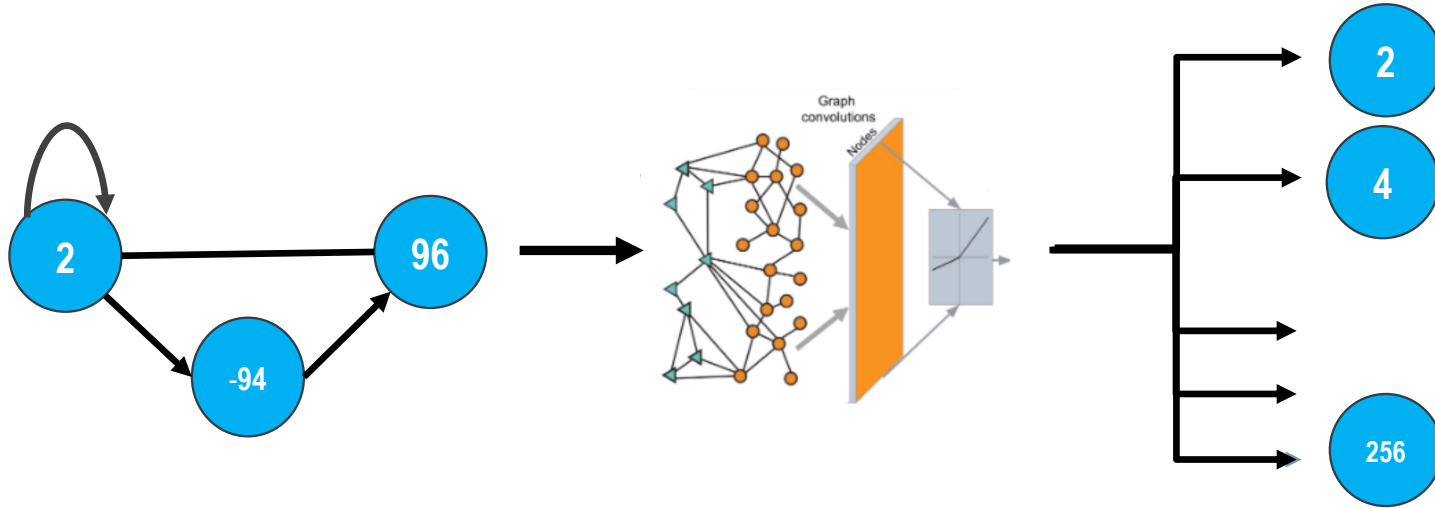
Fully-connected
edges reveal spatial
pattern.

$$\begin{matrix} & \begin{matrix} 2 & 96 & -94 \end{matrix} \\ \begin{matrix} 2 \\ 96 \\ -94 \end{matrix} & \begin{bmatrix} 3 & 0.5 & 1 \\ 0.5 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

Another example



Step 4: Feed into ML



$$\begin{matrix} & 2 & 96 & -94 \\ \begin{matrix} 2 \\ 96 \\ -94 \end{matrix} & \begin{bmatrix} 3 & 0.5 & 1 \\ 0.5 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

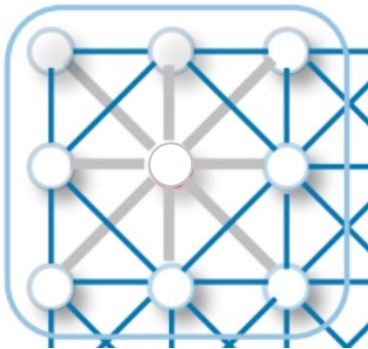
**1000 delta
addresses.**

What ML model



[SGDP '23]
**Message
passing**

Local patterns.
Many iterations.



[Ours]
**Spectral
network**

Global structure.
One pass.



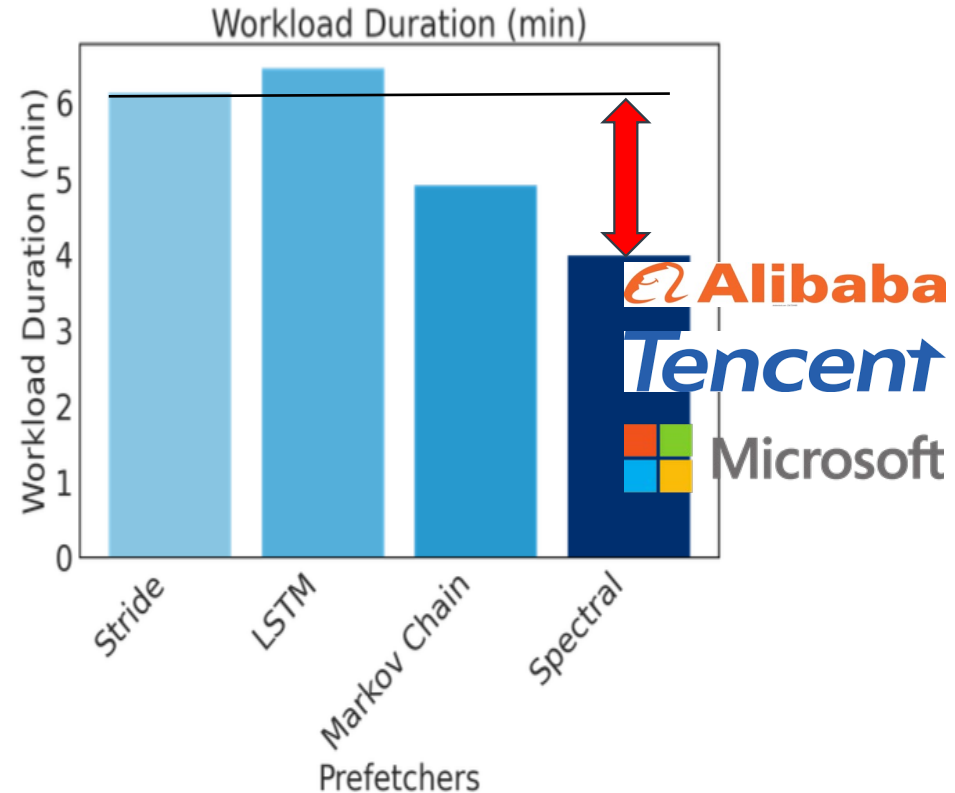
Evaluations

—

Faster access speed

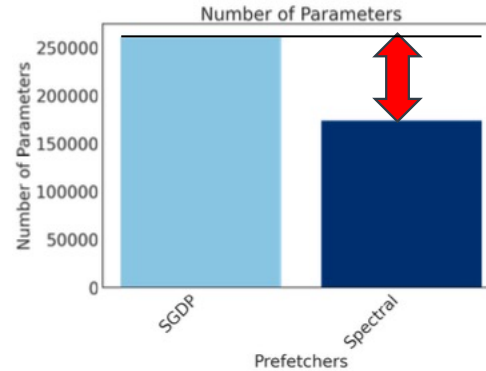
- ❑ Setup: python simulator.
- ❑ Respected arrival time of requests

**Saves 33%
access time!**



Saves memory and time

- Similar hit rate, but much smaller model and faster training and inference than state-of-the-art, [SGDP '23].



**Saves 33.4%
model size.**



**Saves 79%
training time.**

Conclusion

—

Entire pipeline

Raw Address

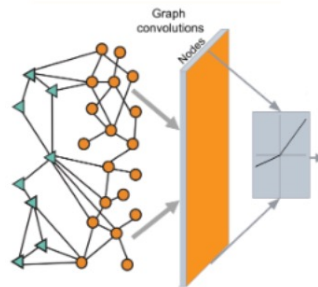
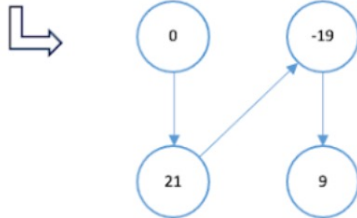


Address Delta

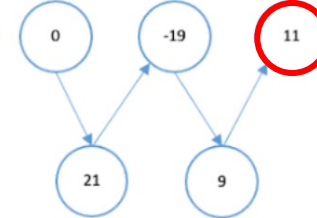


Build a
distilled
graph.

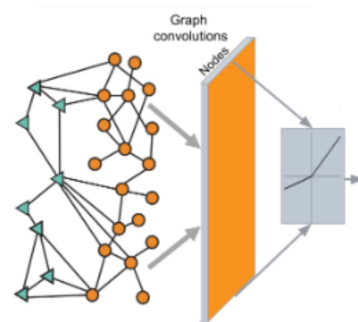
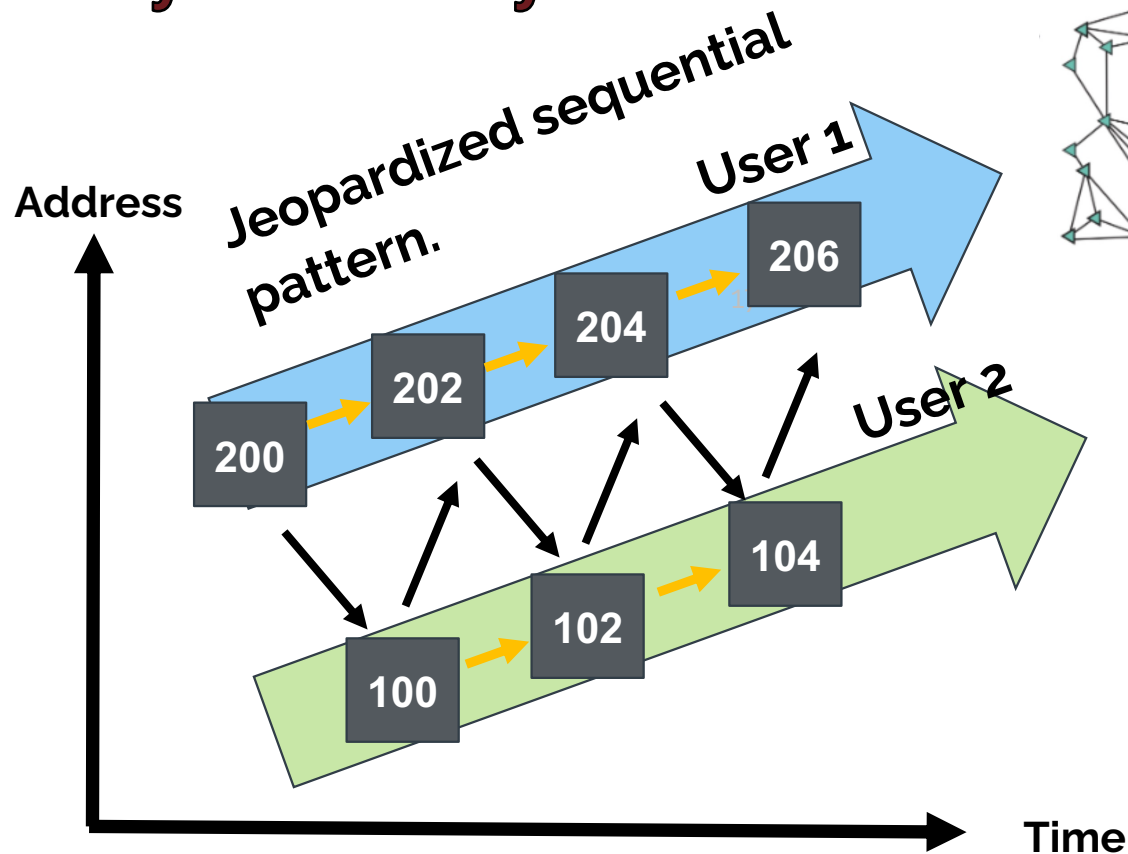
Directed Graph



Predict a
new node.



Key takeaway



Use **graph** to detect a global pattern.

Thank you!

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Backup Slides

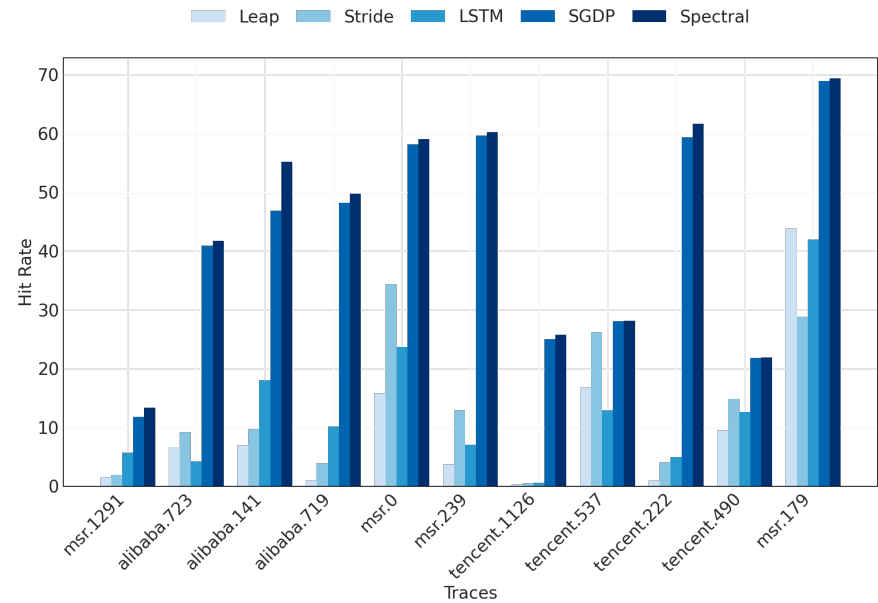
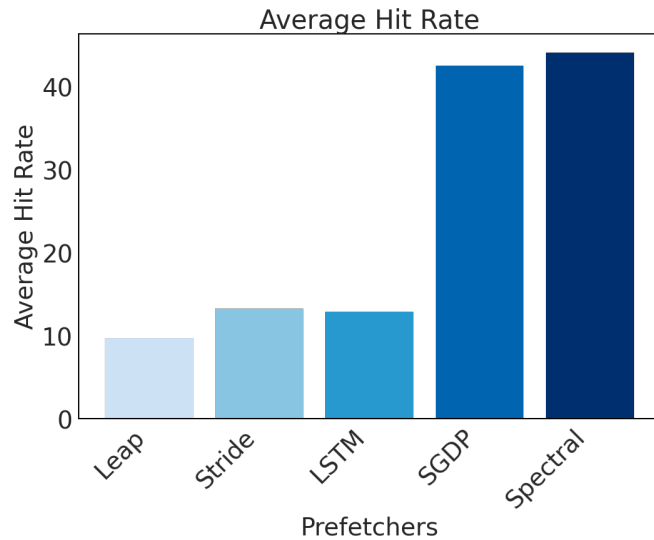
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Is graph building too costly

- ❑ Only putting 32 integers into a matrix.
- ❑ Main Bottleneck is disk access (20ms), then inference (2ms), then graph building (1ms).
- ❑ Spectral < SGDP < LSTM < transformer.

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

Higher hit rate



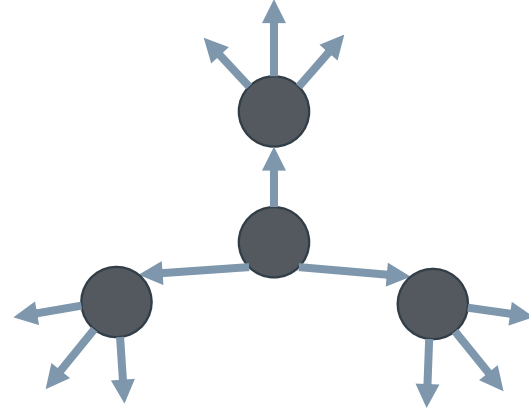
Why spectral method

Message passing

$$m_v^{(t+1)} = \sum_{u \in \mathcal{N}(v)} M^{(t)} \left(h_v^{(t)}, h_u^{(t)}, e_{uv} \right)$$

$$h_v^{(t+1)} = U^{(t)} \left(h_v^{(t)}, m_v^{(t+1)} \right)$$

Many
iterations



Spectral method

$$H^{(2)} = \sigma_2 \left(\hat{A} \sigma_1 \left(\hat{A} X W^{(0)} \right) W^{(1)} \right)$$

One go

