

A Hierarchical Neural Model of Data Prefetching

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Introduction

- Data prefetching presents two challenges to machine learning:
 - 1. class explosion problem
 - 2. labeling problem
- Previous Notable Work: Hashemi, delta correlations
- This Paper's Goal: Voyager, delta & address correlations

Introduction

- Data prefetching presents two challenges to machine learning:
 - 1. class explosion problem
 - 2. labeling problem
- 1. decompose address prediction into two sub-problems:
 - page prediction
 - offset prediction (offset aliasing problem)
- 2. multi-label training scheme

Problem Formulation

- How should we think of Data Prefetching?

Data prefetching – a probabilistic prediction problem

Output – a probability distribution

- Address Correlation

$$P(\text{Addr} \mid \text{Access}_1, \text{Access}_2, \dots, \text{Access}_t)$$

- Delta Correlation

$$P(\text{Stride}_{t+1} \mid \text{Stride}_t)$$

* output label, input features

Previous Work – Address Correlations

- **STMS**, predict next addr in global memory access stream
 $P(\text{Addr}_{t+1} \mid \text{Addr}_t)$
- **ISB**, predict next addr by same program counter (PC localization)
 $P(\text{Addr}_{\text{PC}} \mid \text{Addr}_t)$
- **Domino**, use previous two addresses
 $P(\text{Addr}_{t+1} \mid \text{Addr}_{t-1}, \text{Addr}_t)$

Previous Work – Delta Correlations

- **IP stride prefetcher**, use a per-PC output (PC localization)

$$P(\text{Stride}_{\text{PC}} \mid \text{Stride}_t)$$

- **VLDP prefetcher**, use history of past deltas

$$P(\text{Stride}_{t+1} \mid \text{Stride}_{t0}, \text{Stride}_{t1}, \dots, \text{Stride}_{tn})$$

- **Hashemi neural prefetcher**, use history length of l past deltas

$$P(\text{Stride}_{t+1} \mid \text{Stride}_{t-l}, \text{Stride}_{t-l+1}, \dots, \text{Stride}_t)$$

Voyager Approach

- Hierarchical Neural Structure:
 - one part predicting page addresses;
 - other part predicting offsets.
 - Vocabulary: includes both addresses and deltas.
 - Multi-label training scheme:
 - instead of predicting the next address in the global address stream
 - predict the most predictable address from multiple possible labels
- * A neural network's vocabulary is the set of words that the model can admit as input and can produce as output.

Voyager Approach

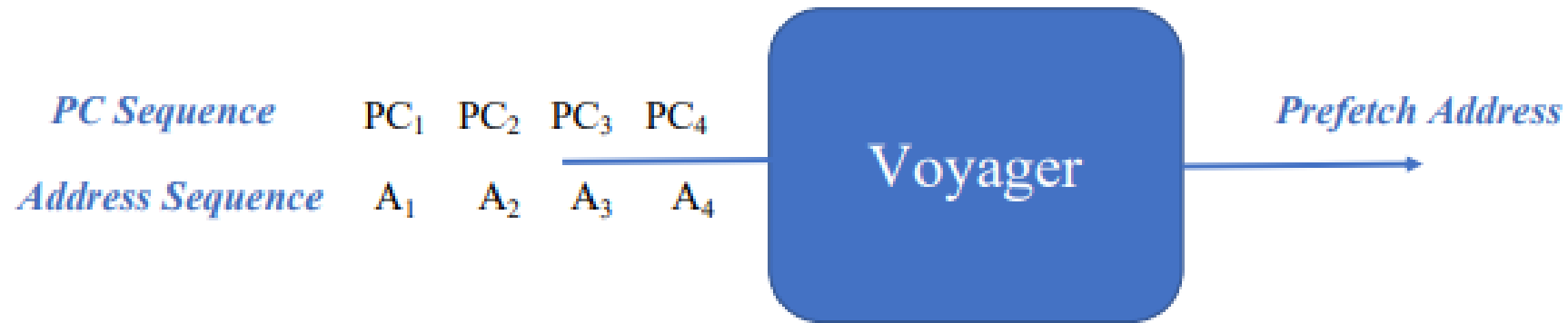


Figure 1: Overview of Voyager.

Voyager Approach

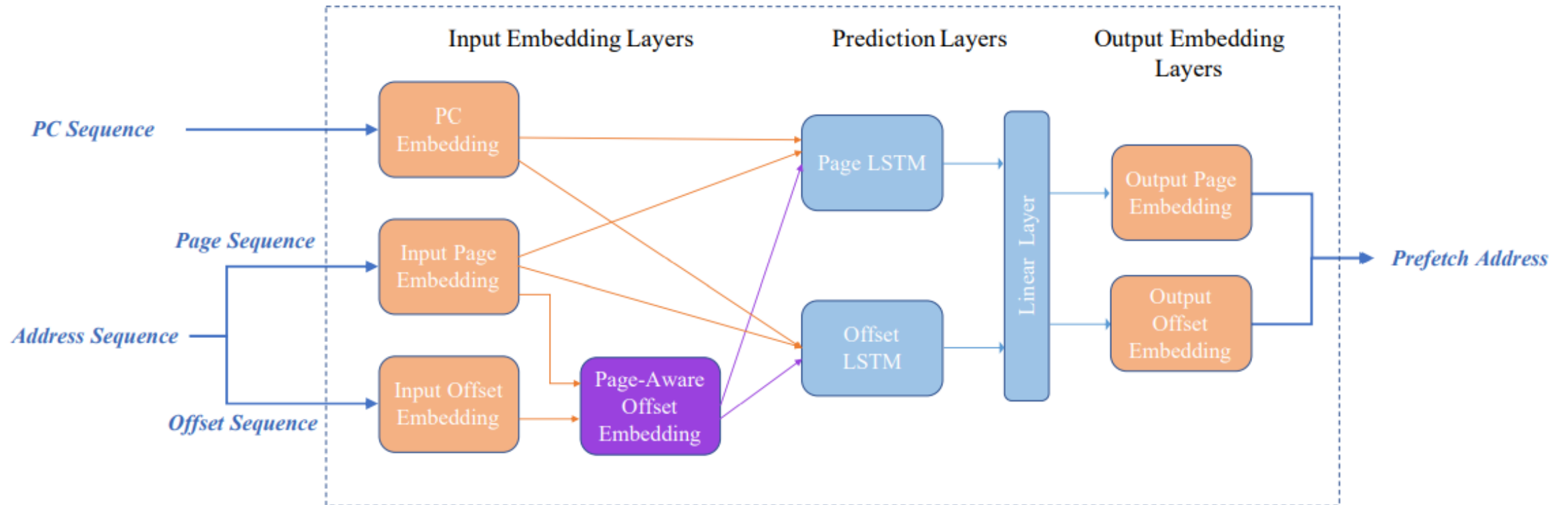


Figure 2: Voyager's Model Architecture.

Results

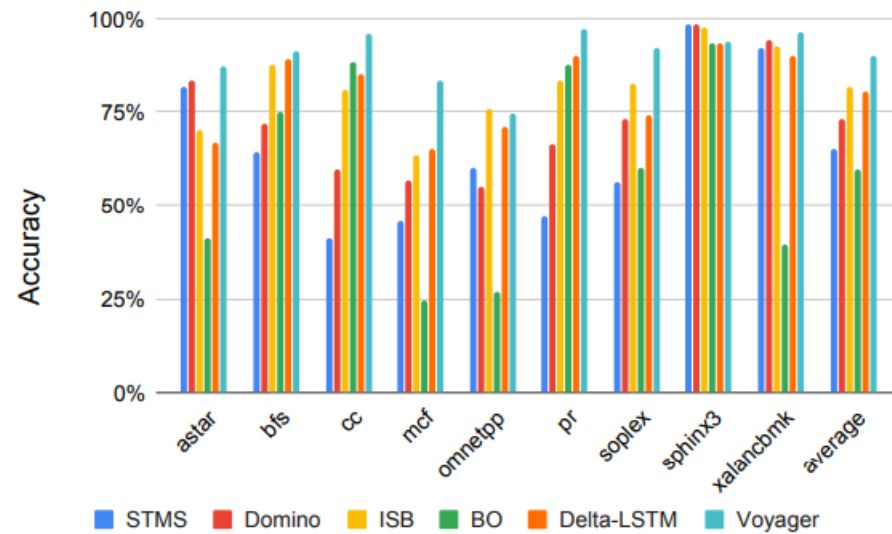


Figure 5: Accuracy.

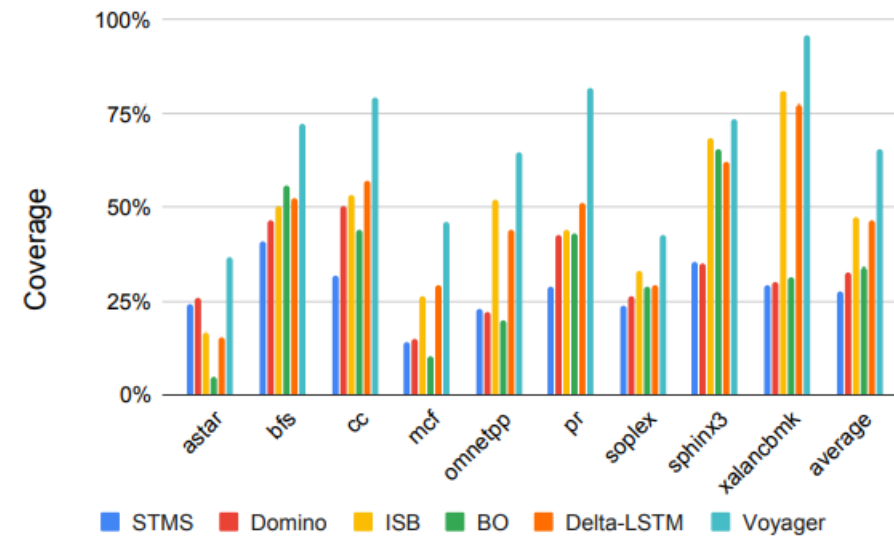


Figure 6: Coverage.

Results

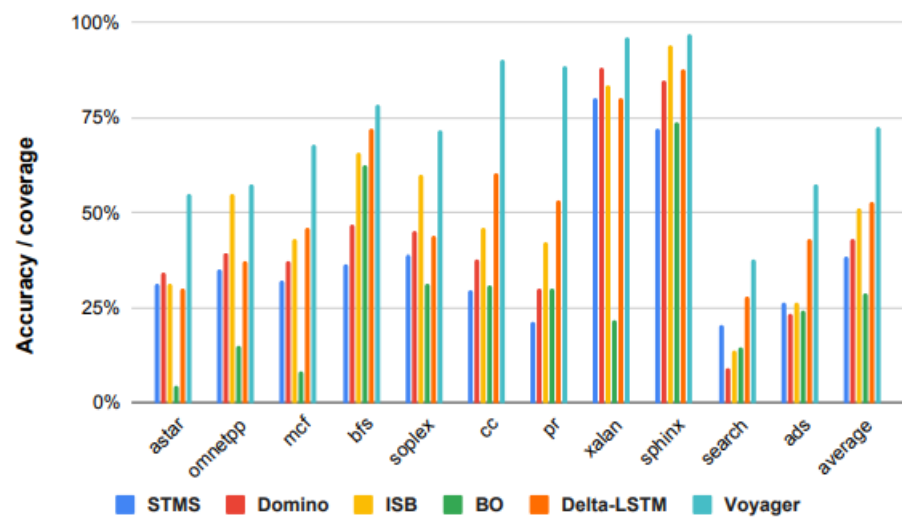


Figure 7: Unified accuracy/coverage, including Google's *search* and *ads*.

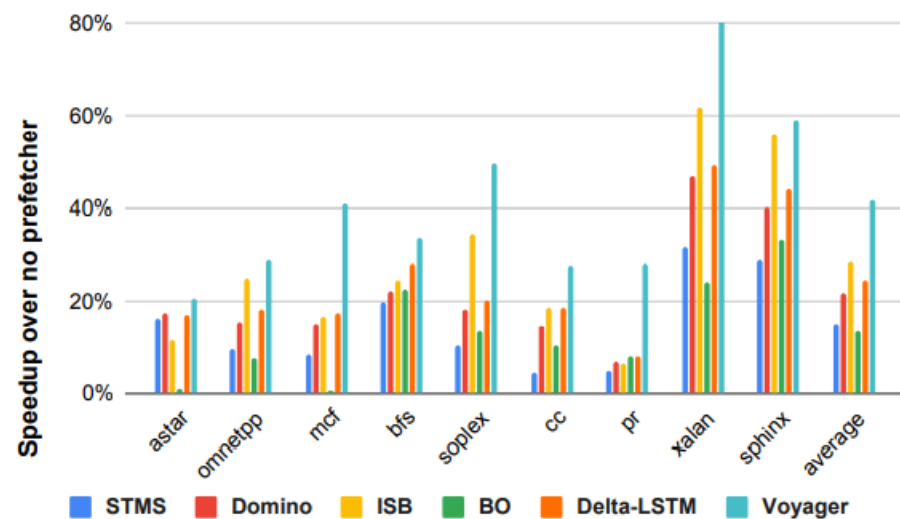


Figure 8: IPC.

Results

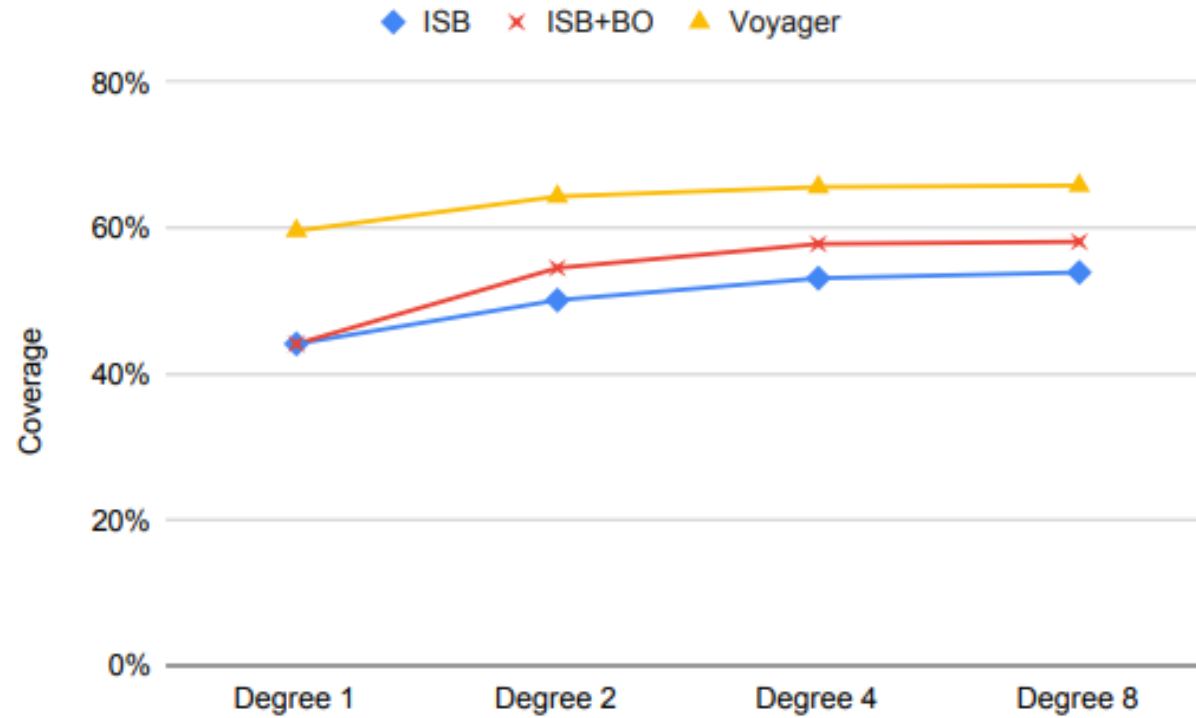


Figure 9: Sensitivity to Prefetch degree.

Results

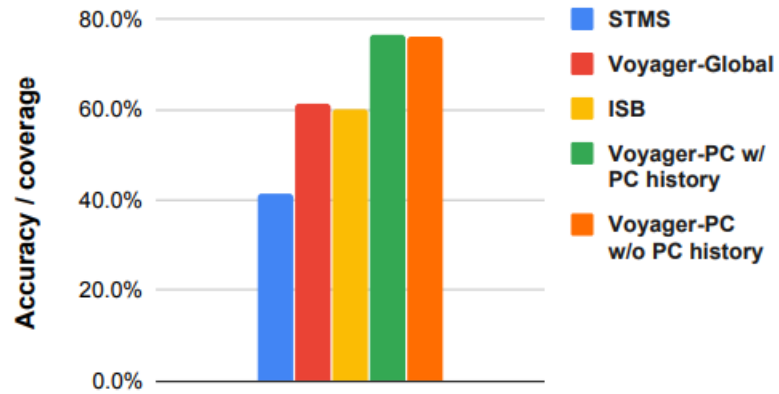


Figure 12: Comparison of different features. Voyager benefits from using data address history as a feature.

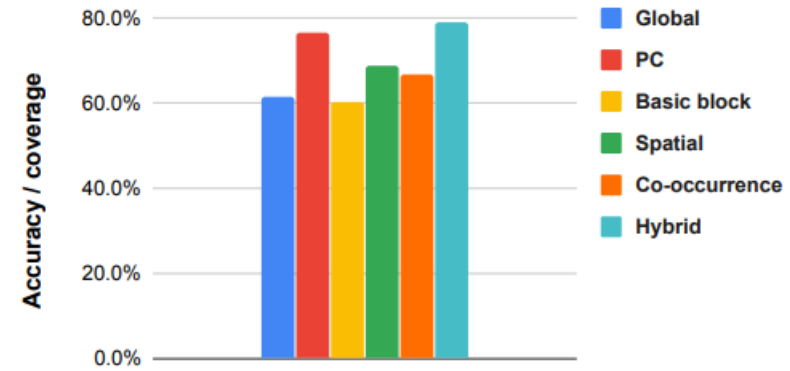


Figure 15: Comparison of different labeling schemes.

Discussion

- Is there anything that we can borrow from this paper into our own project practically?
- If Voyager is too difficult to implement, what about some previous work of prefetching method mentioned within this paper?