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# Pythia

— A Neural Model for Data Prefetching —

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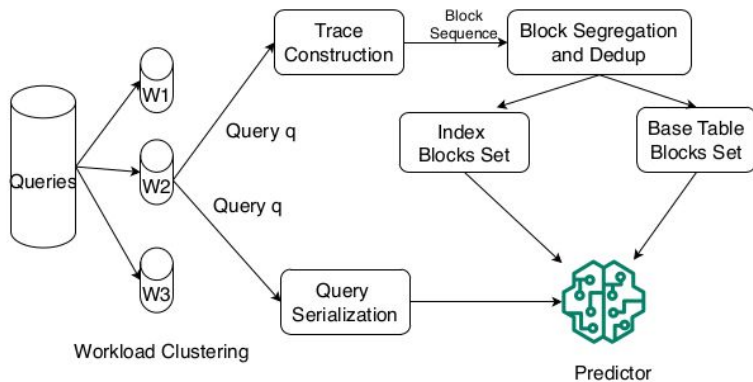
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Akshay Arun Bapat, Saravanan Thirumuruganathan, Nick Koudas

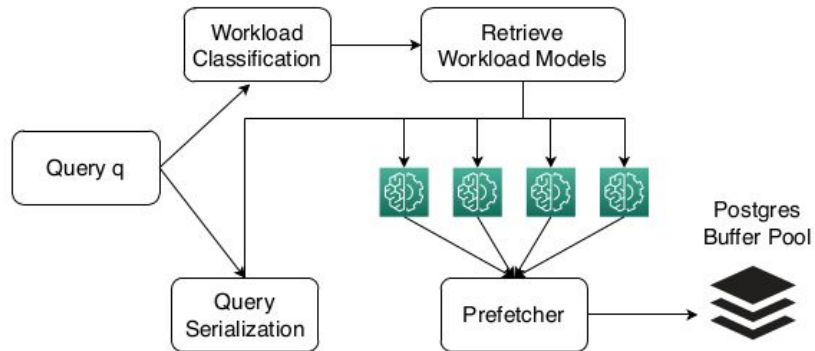
# Motivation

- ML in Databases
  - Cardinality estimation
  - Query optimization
  - Indexing
  - Buffer Manager?
  
- Instance Optimized
  - Offers correlation (predictability) with specific
    - Workload
    - Data distribution
  - Performance boost

# Problem / Goal



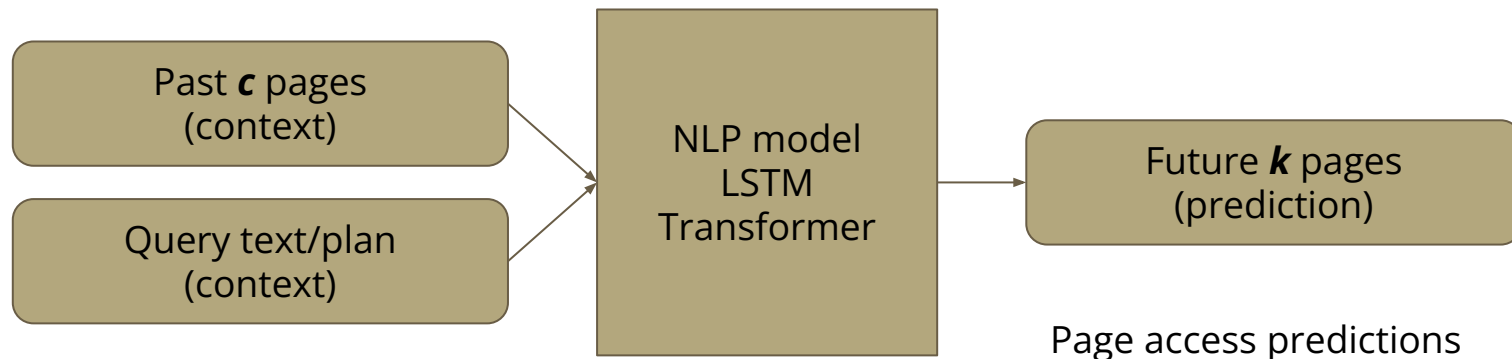
Preprocess



Use

# Page access prediction

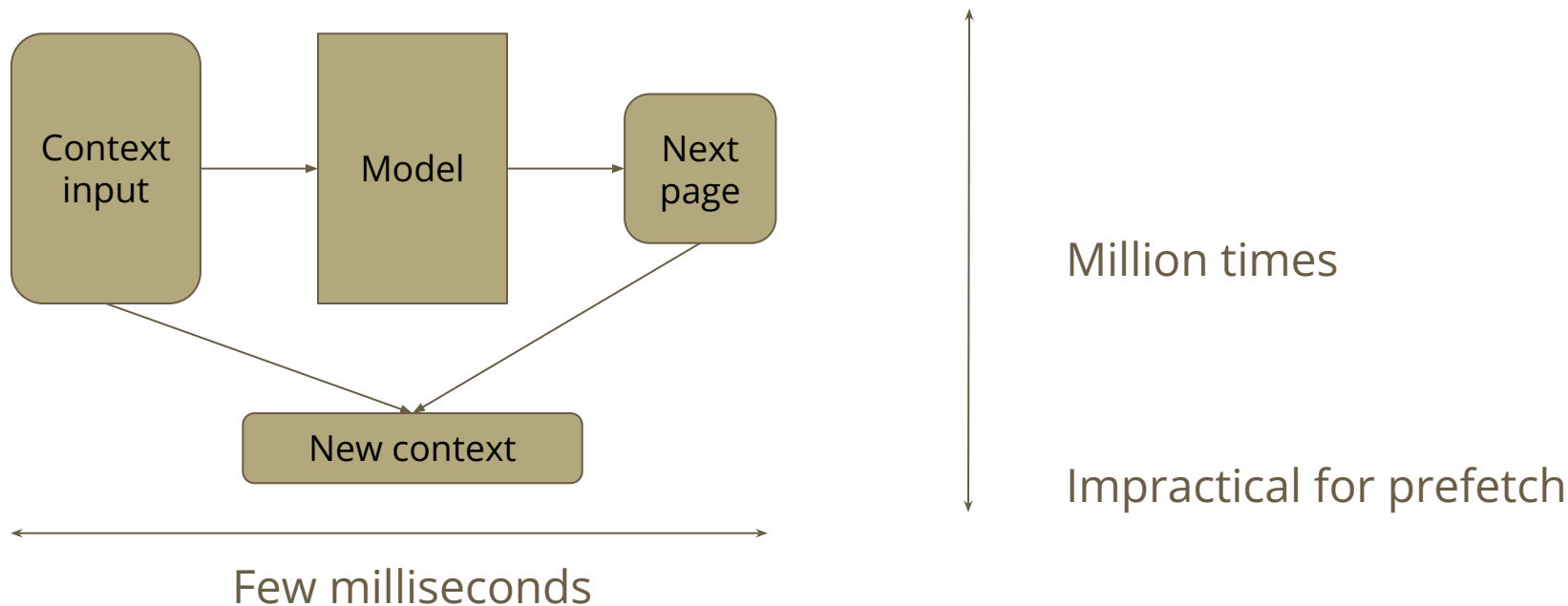
- Predicting access patterns
  - Memory access
  - File access
- Access patterns are a sequence
  - Natural sequence prediction problem
  - NLP models like LSTM and transformers



**Transformers do NOT work!!**

# Transformers do not work!!

- Sequential nature of prediction



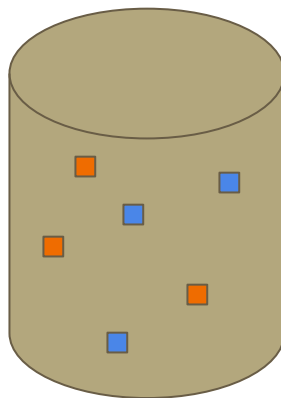
# Transformers do not work!!

- Sensitivity to query predicates

Select \* from Table where column = A

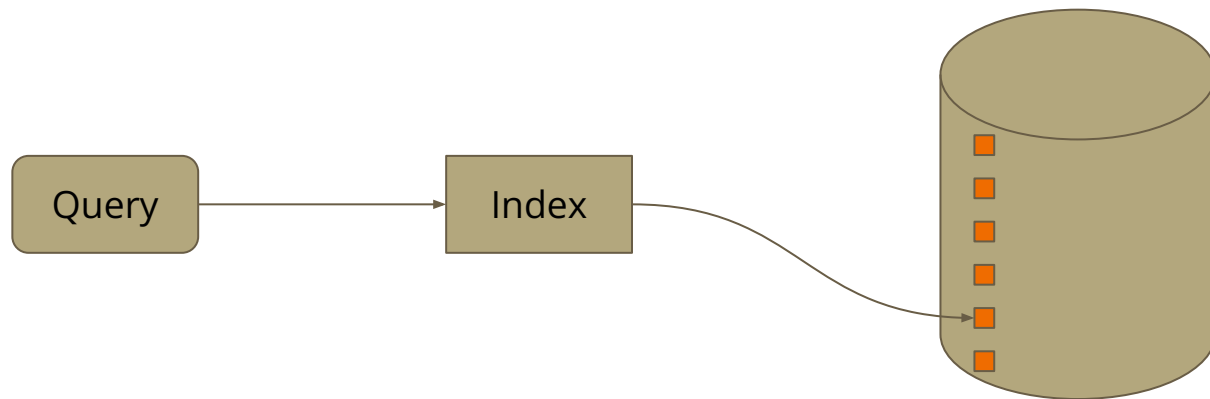


Select \* from Table where column = B



# Transformers do not work!!

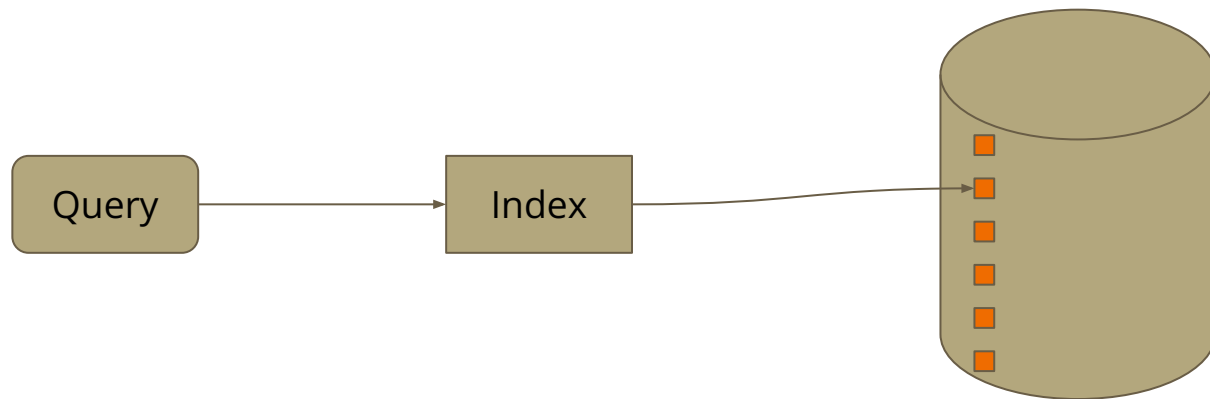
- Irregular block access





# Transformers do not work!!

- Irregular block access

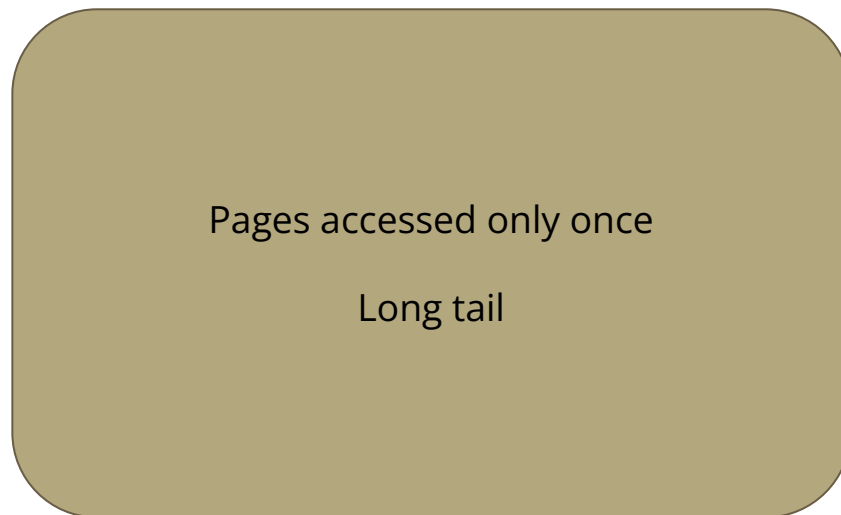


# Transformers do not work!!

- Page access distribution



Pages accessed many times



Pages accessed only once

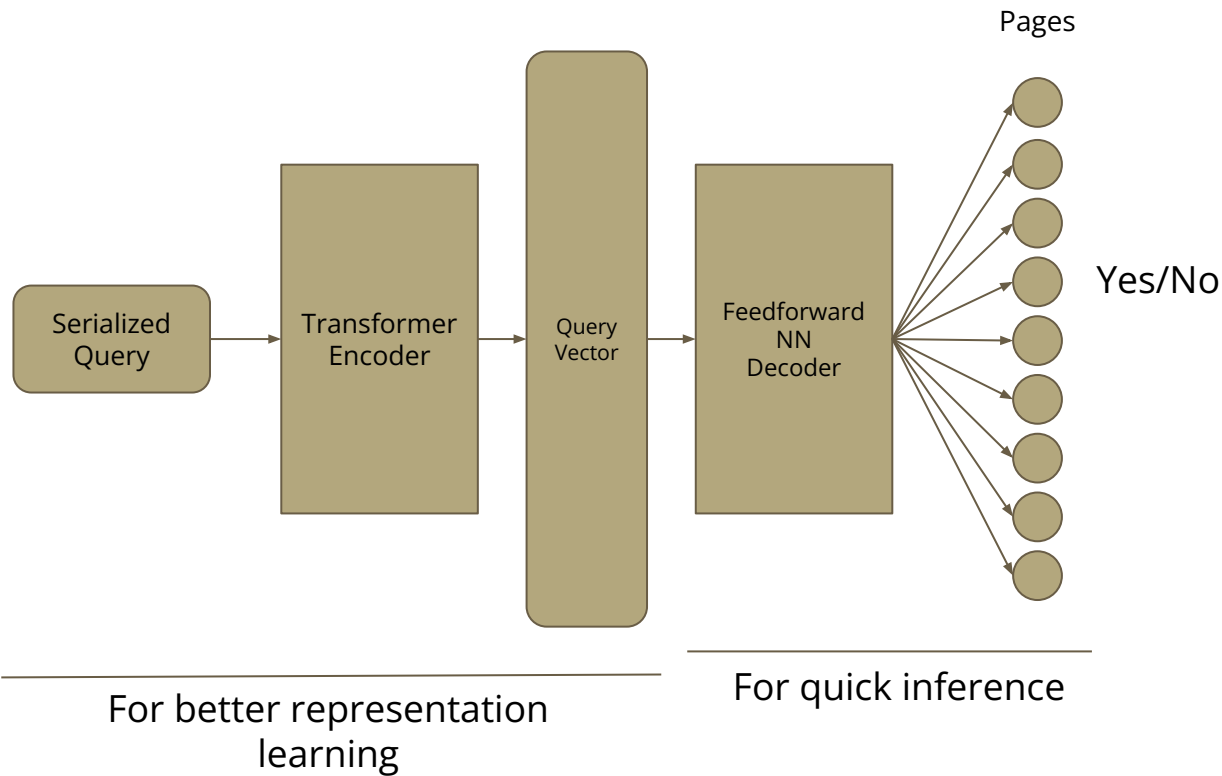
Long tail

Transformers are bad at learning long tail knowledge

**Transformers do not work  
NOW**

**With better and faster models  
Prefetch is possible in the future**

# Hybrid Predictor

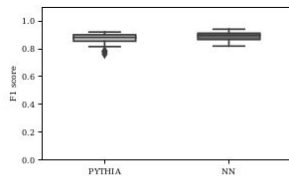


# Experiments

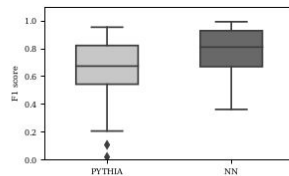
- DSB OLAP benchmark
  - 100GB database
  - Query templates as workloads
- Postgres
  - Enhance postgres to prefetch pages
- Performance metrics
  - F1-score for model accuracy
  - Speedup for prefetch improvement
- Baselines
  - Oracle (ORCL) - knows all future page accesses for a query
  - Nearest neighbour (NN) - knows closest set of page accesses from the training pool

# Experiments

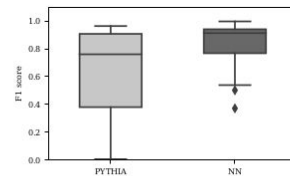
F1-score



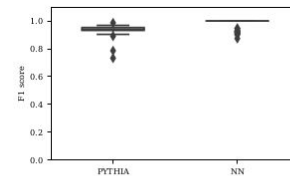
(a) Template 13



(b) Template 18

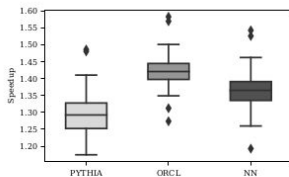


(c) Template 19

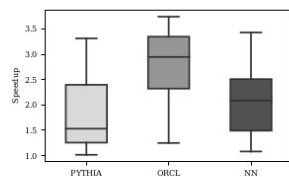


(d) Template 91

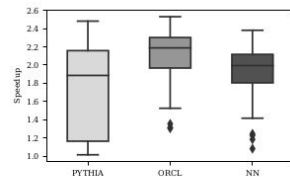
Speedup



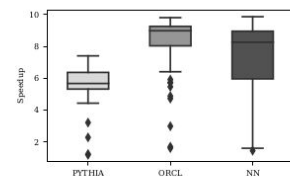
(a) Template 13



(b) Template 18

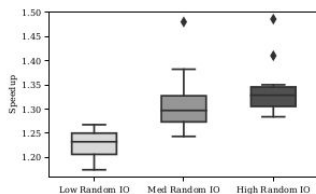


(c) Template 19

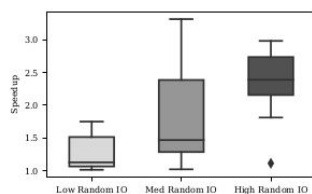


(d) Template 91

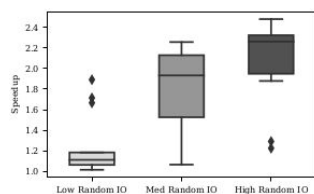
## Impact of random reads



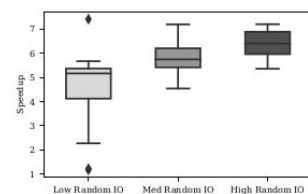
(a) Template 13



(b) Template 18

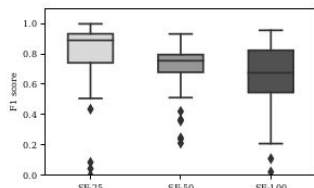


(c) Template 19

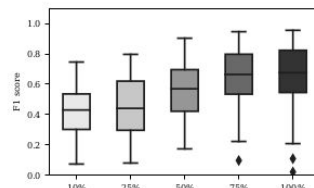


(d) Template 91

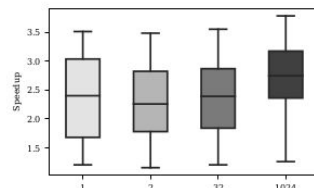
## Impact of miscellaneous factors on template-18



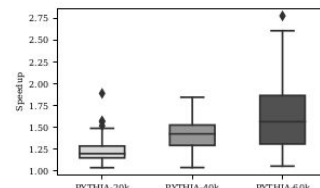
(a) Scale Factor



(b) Training Data Size

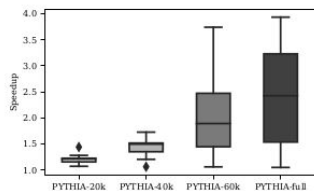


(c) Readahead Size Window

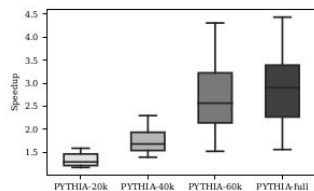


(d) Speedup when predicting top- $k$  frequent pages

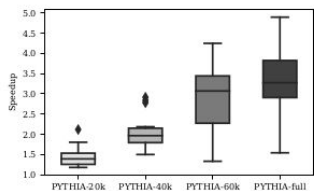
## Impact of concurrent queries on template-18



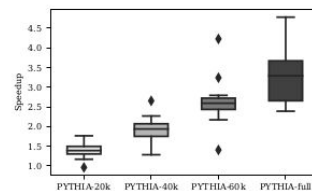
(a) 2 concurrent queries



(b) 3 concurrent queries



(c) 4 concurrent queries



(d) 5 concurrent queries

# Conclusion

- Page access patterns are predictable
- Prefetching provides high performance benefits
- Integration with Postgres for experimental evaluation



**Questions?**