Apollo:Learning Database Query **Correlations for Predictive Caching**

Brad Glasbergen, Khuzaima Daudjee, Michael Abebe, Scott Foggo, Anil Pacaci



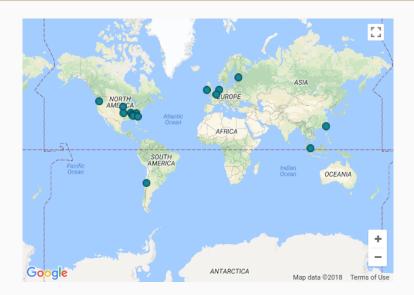


@bglasber

Simple Web Application Architecture



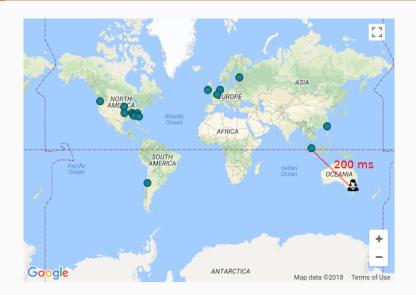
Worldwide Client/Data Center Distribution



Worldwide Client/Data Center Distribution



Worldwide Client/Data Center Distribution



Latency Effects on Clients

Increased latency reduces user engagement, and consequently revenue!

Schurman et al., "Performance Related Changes and Their User Impact". *Velocity*, 2009.

Edge Caching (Content Delivery Networks)



Worldwide Client/Edge Node Distribution

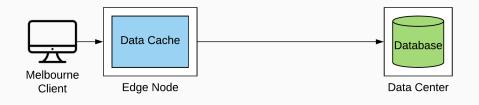


Map of metros where at least one Edge node (GGC) is present.

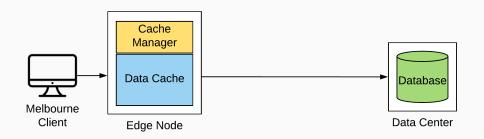
Content Delivery Networks — A Silver Bullet?

• Limited support for non-static data!

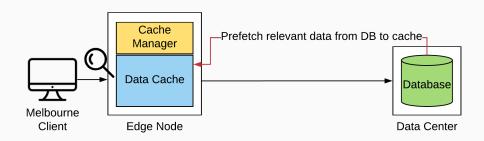
Static Data Edge Cache Architecture



Extending Edge Cache Support



Extending Edge Cache Support



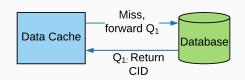
Dynamic Data Requests (TPC-W Benchmark)

```
1. SELECT C_{-}ID FROM CUSTOMER WHERE C_{-}UNAME = @C_{-}UN and C_{-}PASSWD = @C_{-}PAS
```

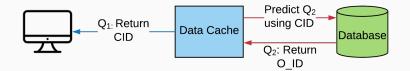


 $\textbf{Q_1}\!\!:$ Look up customer ID $\textbf{Q_2}\!\!:$ Look up last order for customer ID

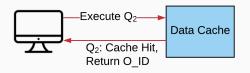




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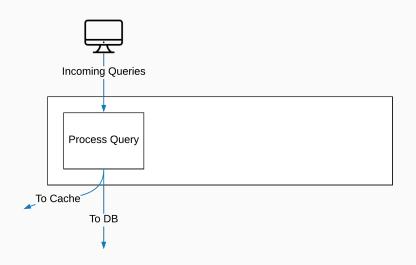
- Uses online learning to discover client query patterns.
- Predictively executes and caches query results using these patterns to reduce client response time.
- Employs a computationally efficient means of managing updates to cached data.

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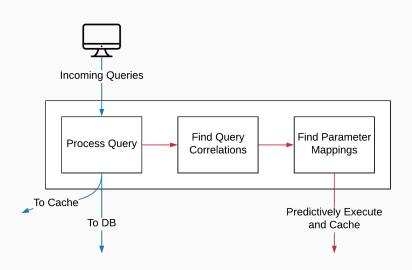
- 1. Predictive Query Model
- 2. Cache Management
- 3. Results

Predictive Query Model

Apollo Overview



Apollo Overview



A Query Submission

```
SELECT C_{-}ID FROM CUSTOMER WHERE C_{-}UNAME = 'Alice' and C_{-}PASSWD = 'pass'
```

SELECT MAX(O_ID) FROM ORDERS WHERE $O_C_ID = 3$

Query Templates

Two query instances, Q_1 , Q_2 share the same query template if they have the same query text modulo *parameterizable* constants.

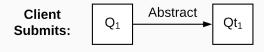
Abstracting Query Instances to Query Templates

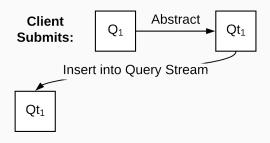
```
SELECT C_ID FROM CUSTOMER WHERE C_UNAME = ? and C_PASSWD = ?
```

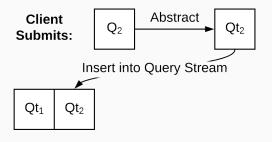
SELECT MAX(O_ID) FROM ORDERS WHERE O_C_ID = ?

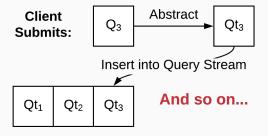
Client Submits:

 Q_1

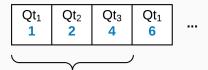


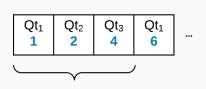


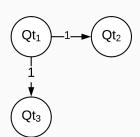


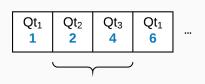


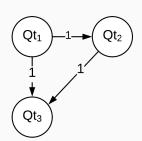


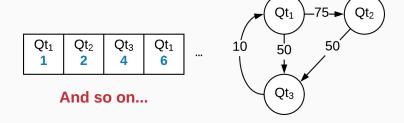




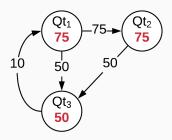




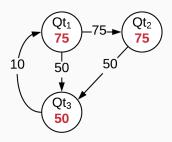




Query Transition Graph



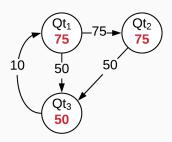
Query Transition Graph



Probability of seeing Qt_2 within sliding window after we've seen Qt_1 :

$$P(Qt_2|Qt_1; T \leq \Delta t) = \frac{75}{75} = 1$$

Query Transition Graph



Probability of seeing Qt_2 within sliding window after we've seen Qt_1 :

$$P(Qt_2|Qt_1; T \le \Delta t) = \frac{75}{75} = 1$$

 $P(Qt_1|Qt_3; T \le \Delta t) = \frac{10}{50} = \frac{1}{5}$

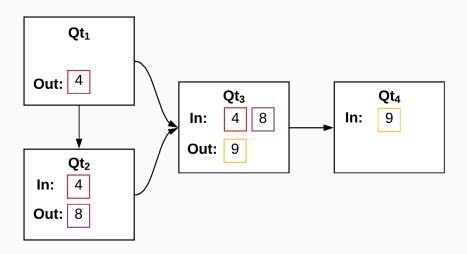
```
SELECT C_ID FROM CUSTOMER WHERE C_UNAME = ? and C_PASSWD = ?
```

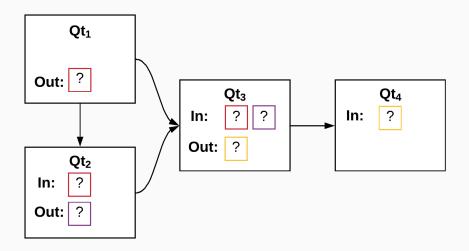
SELECT MAX(O_ID) FROM ORDERS WHERE $O_LC_ID = ?$

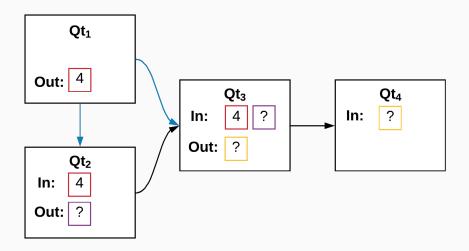
```
3
SELECT C_ID FROM CUSTOMER WHERE C_UNAME = ? and C_PASSWD = ?
SELECT MAX(O_ID) FROM ORDERS WHERE O_C_ID = ?
```

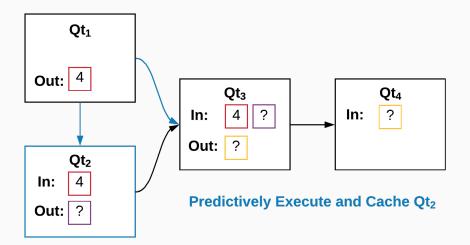
```
SELECT C_ID FROM CUSTOMER WHERE C_UNAME = ? and C_PASSWD = ? SELECT \ MAX(O_ID) \ FROM \ ORDERS \ WHERE \ O_C_ID = ?
```

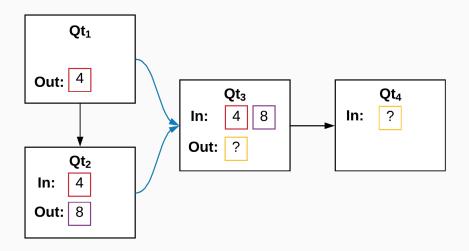
Dependency Graph

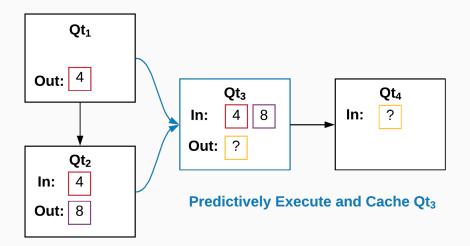


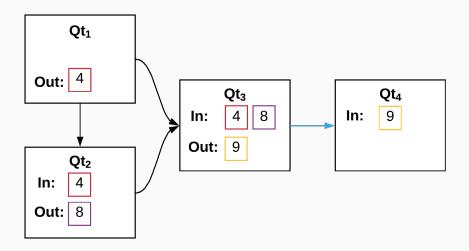


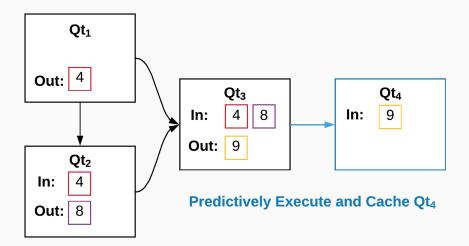




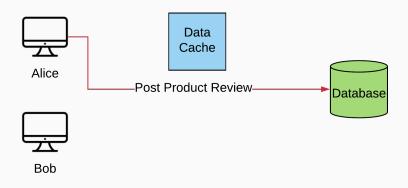


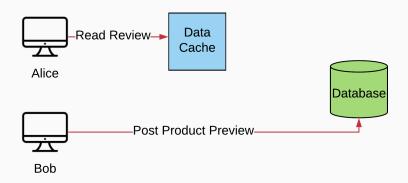


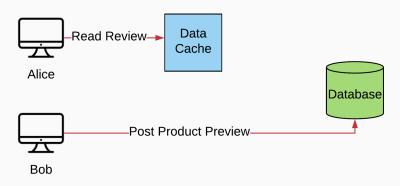




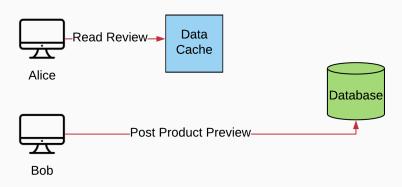
Cache Management







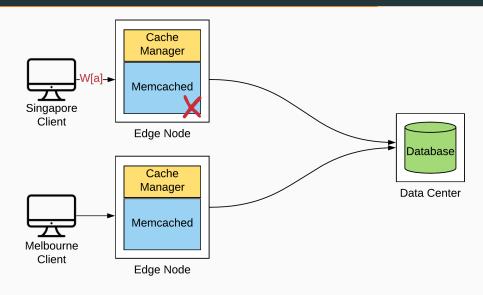
Alice should see her own order, but does not care about Bob's!



Alice should see her own order, but does not care about Bob's!

- Improves cache performance, client-centric model
- Support for reading latest data if needed

Benefits of Client-Centric Caching



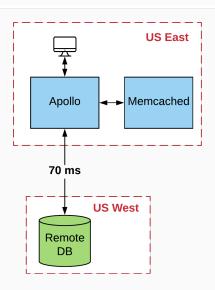
Benefits of Client-Centric Caching

- Can predict whether a prefetched query result will be used before invalidation!
- Reloading queries upon invalidation

See paper for details!

Results

Experiment Configuration



Experiment Configuration

Three configurations:

- Apollo configuration: as described in prior sections.
- Memcached configuration: LRU cache Apollo with predictive features turned off
- Fido configuration: Use Fido predictive engine instead of Apollo's predictive features!

Fido



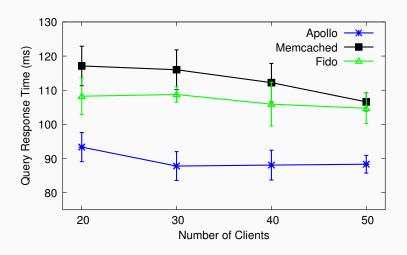
Fido



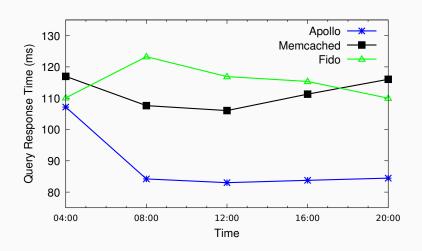
Fido

- Query instance based predictions, instead of query templates.
- Prefix length: 3, Suffix Length: 2
- Requires offline training (Supplied 40 minutes of data).

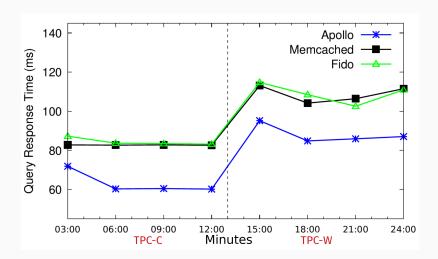
TPC-W Results



Learning Over Time



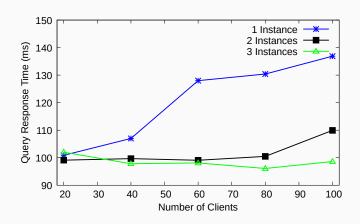
Workload Change



Predictive Caching Works

- Online learning enables workload pattern discovery and adapts to client behaviour changes
- Predictive caching is an effective tool to reduce application latency

Multiple Apollo Instances



Parameter Settings (TPC-W)

Parameter	Setting
Window Width	15s
Correlation Threshold	0.99
Reload Threshold	0
Cache Size	5% of DB
Cache Size	5% of DE

Related Work (Systems)

Similarities and Differences A	Among	Related	Work
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System	Similarities	Differences
Scalpel	Prefetching via templates	Offline trainingWrite HandlingClient-sideQuery Rewriting
Fido	PrefetchingServer-side/middleware	 Offline training Query Instances Write Handling

K. Amiri, S. Park, R. Tewari, and S. Padmanabhan.

Dbproxy: a dynamic data cache for web applications.

In Proceedings 19th International Conference on Data Engineering (Cat. No.03CH37405), pages 821-831, March 2003.



E. Schurman and J. Brutlag.

Performance related changes and their user impact. Velocity, 2009.