# On Sample-Path Staleness in Lazy Data Replication

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April 29, 2015

- Introduction
- Staleness Formulation
- Staleness Cost
- Optimality
- Applications
- Conclusion

#### **Introduction**

- Highly-dynamic content
  - News
  - Weather
  - Road conditions
- An increasing number of applications need to maintain local copies of remote data sources
  - Search engines
  - Mash-up applications
  - Distributed caching
- Copies need to be synchronized constantly
  - To provide reliable services

#### **Introduction**

- Push-based policy
  - Sources send the update information to replicas
  - Requires the cooperation of sources
  - Hard to scale
- Pull-based policy
  - Replica retrievals the content explicitly
  - Scalable and less costly
  - Leads to staleness
- Need models and mechanisms for analyzing and controlling staleness
  - Previous works mainly consider Poisson updates

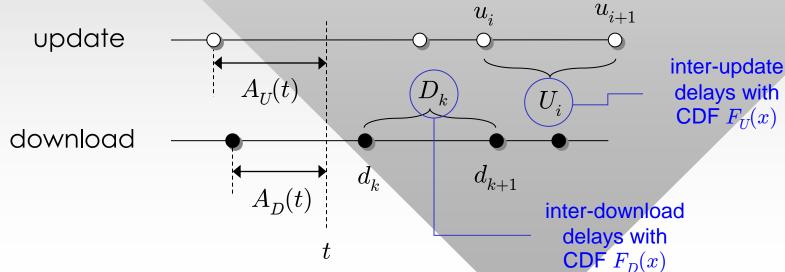
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#### **Staleness Formulation**

Information requests indicated by arrows:

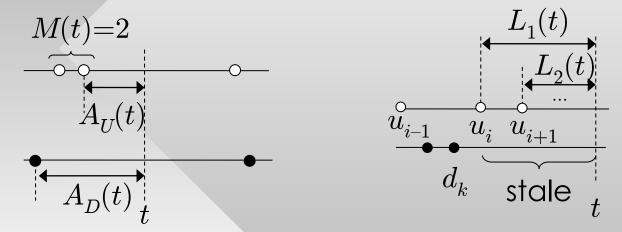


- Model
  - Source experiences random updates via process  $N_{\!U}$
  - Replica periodically downloads the content via process  ${\cal N}_{D}$



### **Metrics**

- M(t): the number of updates missing from the replica
  - E[M(t)]: the expected number of missing update



Backward delays to each unseen update

$$L_1(t) > L_2(t) > ... > L_{M(t)}(t)$$

- Apply weight function w(x) to each lag
  - Maps staleness lags to actual cost

#### **Metrics**

Two different cost metrics:

$$\eta(t) = \begin{cases} w(L_1(t)) & M(t) > 0 \\ 0 & \text{otherwise} \end{cases} \quad \rho(t) = \begin{cases} \sum_{i=1}^{M(t)} w(L_i(t)) & M(t) > 0 \\ 0 & \text{otherwise} \end{cases}$$
 Both  $\rho(t)$  and  $\eta(t)$  generalize metrics in previous work

- - $\eta(t)$  becomes staleness and age [Cho 2000] with w(x)=1 and w(x) = x
  - $\rho(t)$  with w(x)=1 and w(x)=x lead to  $\mathit{blur}$  [Denev 2009] and addictive age [Ling 2004]
- Both  $\rho(t)$  and  $\eta(t)$  are random variables
  - Using expectation requires multiple sample-paths
  - One single sequence available in practice (sample-path)

$$\bar{\eta} = \lim_{T \to \infty} \frac{1}{T} \int_0^T \eta(t) dt$$
 and  $\bar{\rho} = \lim_{T \to \infty} \frac{1}{T} \int_0^T \rho(t) dt$ .

### **Staleness Cost**

- Phase-lock problem prevents us from getting the solution of the above two metrics
  - Constant update interval with interval 1
  - Constant download interval with interval 2
  - The staleness metrics depends on their initial states
- To avoid phase-lock cases, we propose age independence assumption:
  - Random query time of consumers  $Q_T$ : uniform in [0,T]
  - Two points processes  $N_U$  and  $N_U$  are called age-independent if  $\forall x,y>0$

$$\lim_{T \to \infty} P(A_D(Q_T) < x | A_U(Q_T) = y) = G_D(x)$$

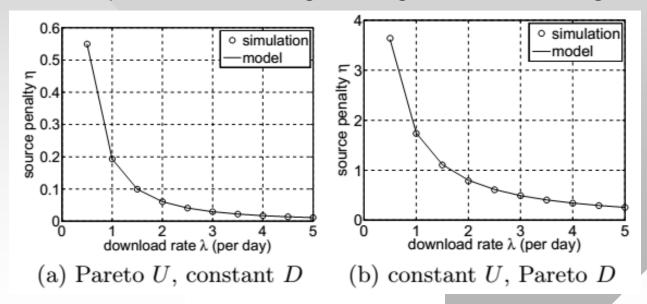
sample path equilibrium distribution of  ${\cal A}_{\cal D}$ 

## **Staleness Cost**

Theorem 1: Source penalty: update rate

$$\bar{\eta} = \lambda \mu \int_0^\infty \bar{F}_U(y) \int_0^\infty w(x) \bar{F}_D(x+y) dx dy$$

- Matches previous results with exponential update
- When  $\lambda = \mu$ , left scenario gives age 1.5 hours; right 19 hours



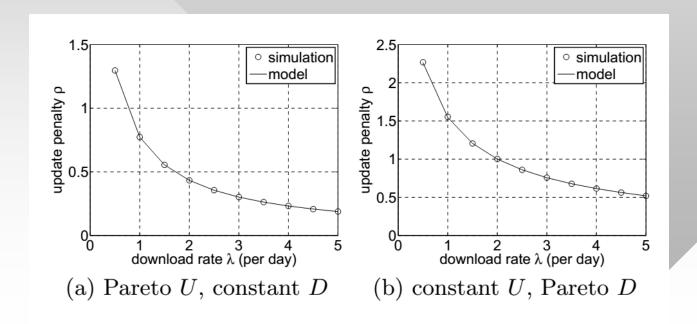
$$w(x) = x, \mu = 2$$

# **Staleness Cost**

Theorem 2: Update penalty:

$$\bar{\rho} = \mu E[w_2(A_D)] = \lambda \mu E[w_3(D)] \quad w_n(x) = \int_0^x w_{n-1}(y) dy$$

- Matches previous results with exponential update
- Allows decaying function w(x) = 1/(1+x)



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# **Optimality**

- With the same download rate  $\lambda$ , what distribution of synchronization intervals  $F_D(x)$  is best?
- <u>Definition 7</u>: Variable X is stochastically larger than Y in second order, i.e.,  $X \ge_{st}^2 Y$ , if

$$\int_0^x \bar{F}_X(y)dy \ge \int_0^x \bar{F}_Y(y)dy \text{ for all } x \ge 0$$

• Theorem 7: For a given download rate  $\lambda$  and fixed update process  $N_U$ , both  $\bar{\eta}$  and  $\bar{\rho}$  decreases if download delays become stochastically larger in second order

# **Best Download Strategy**

- Similarly, for a given update rate  $\mu$  and fixed download process  $N_D$ , freshness increases if inter-update delays become stochastically smaller in second order
- <u>Lemma 3</u>: For a given mean, a constant stochastically dominates all other random variables in second order
- Corollary 1: Constant inter-synchronization delays are optimal under both  $\eta$  and  $\rho$ , all suitable weights w(x), and all update processes  $N_U$

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#### Real-Life Update Process

- The most frequently modified article in Wikipedia
  - "George W. Bush" with 44,296 updates in 10 years
  - (a) Pareto tail  $(1 + x/\beta)^{-\alpha}$  with  $\alpha = 1.4, \beta = 0.93$
  - (b) Long-range dependence
     with Hurst parameter 0.81
  - (c) Non stationary
  - (d) Comparison between
     Poisson and real updates

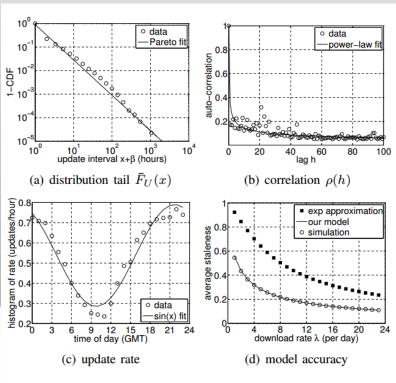
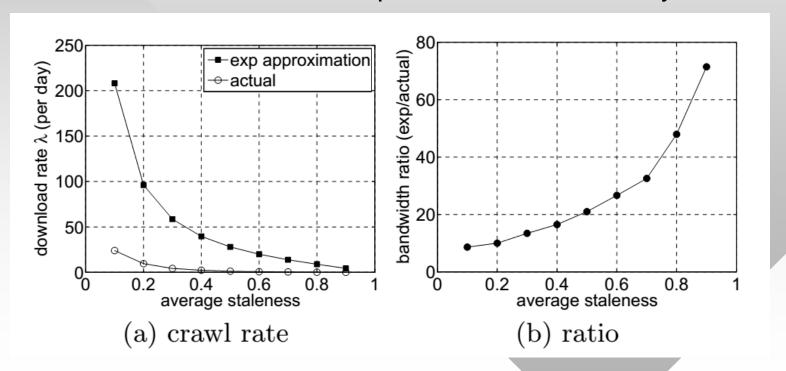


Fig. 7. George W. Bush page dynamics.

### **Bandwidth Estimation**

- Comparison of our results with previous work
  - Apply both models to George W. Bush
  - Poisson: 20% staleness requires 95 downloads/day
  - Real: 20% staleness requires 8 downloads/day



## Aggregation (Many-to-One)

Resource allocation

probability that an incoming query quest i

update rate of source i subject to  $\sum_{i=1}^{n} \lambda_i \leq \Lambda$ 

download rate of source i

- Using source penalty  $\overline{\eta}$ :
  - Page starvation exists: a source i will never be synchronized when total bandwidth  $\Lambda$  is relatively small compared to the source update rate  $\mu_i$
- Theorem 4: Assume  $q_i\mu_i>q_j\mu_j>0$  and constant download delay, optimal solution using  $\bar{\rho}$  guarantees that  $\lambda_i > \lambda_j > 0$ 
  - No page starves  $\lambda_i = \Lambda \frac{\sqrt{q_i \mu_i}}{\sum_{i=1}^{M} \sqrt{q_i \mu_i}}$ .

### **Load-Balancing (One-to-Many)**

- Single source multiple replicas
  - The goal is to deduce the expected penalty afforded by the freshest member of all m replicas
  - Each replica has rate  $\lambda/m$
  - Compare with single replica with rate  $\lambda$
- The staleness at different replicas is no longer independent
  - Updates at the source make all copies outdated
  - The entire collection of replicas can be replaced by a single replica that that refresh pattern  $N_D^*$ , which is the superposition of all point processes  $\{N_D^i\}_{i=1}^m$

#### **Conclusion**

- We proposed a novel framework for modeling staleness metrics under general update/download processes
- We established that constant inter-refresh intervals were optimal for all considered cases
- Finally, we consider a family of related problem stemming from  $1 \times m$  and  $M \times 1$  replication, showing that they can be solved from the preceding results

Questions?