Mace, J. *et al.*, "2DFQ: Two-Dimensional Fair Queuing for Multi-Tenant Cloud Services," *Proc. of ACM SIGCOMM '16*, 46(4):144-159, Aug. 2016.

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# Introduction

- Server has limited capacity
- Requests by clients are queued
- Crucial to provide resource isolation to ensure that a single tenant cannot get more than its fair share of resources

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## **Differences From Traditional Queueing**

- Resource concurrency: threads in shared process handle requests concurrently
- Large cost variance: request costs vary by at least 10000.
- Unknown and unpredictable resource costs: Unknown at schedule time and difficult to estimate
  - Wrong estimates are penalized heavily
  - Network schedulers rely on this information

# Bursty Schedules from WFQ and WF<sup>2</sup>Q

- Both methods below observe long term fairness
- Goal is to achieve smoother schedules and achieve fairness in a smaller time scale



# Key Insights

- Separate requests with different costs across different worker threads
- Use cost estimation to locate unpredictable requests and separate them from predictable requests
- 2D = across both time and different threads
- Desired traits .
  - Work conserving
  - Achieve fairness over shorter time periods

# High Request Cost Variability



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Costs can vary by 4 orders of magnitude Bursty scheduling adversely affects tenants with small requests • Get serviced in high throughput bursts than evenly paced over time

# **Unknown Request Costs**



Request rates can be unpredictable and vary by around 1.5 orders of magnitude

# **Unknown Request Costs**



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- Each API has tenants using it in both stable and unpredictable ways Can use request cost estimations, but these fall apart for more unpredictable tenants •

# **Unknown Request Costs**

- Incorrect cost estimates can lead to bursty schedules
- Expensive request gets predicted as an inexpensive request
  - Blocks worker thread for longer than expected
  - Scheduler can incorrectly schedule up to N (number of threads) requests
- Insight: give good service to predictable tenants

# WFQ

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$$S(r_{f}^{j}) = max\{v(A(r_{f}^{j})), F(r_{f}^{j-1})\}$$

 $F(r_f^j) = S(r_f^j) + I_f^j / \Phi_f$ 

- A(r<sub>f</sub><sup>j</sup>) = walltime arrival time of packet
- $F(r_f^j)$  = virtual finish time
- I<sup>j</sup><sub>f</sub> = size of request
- Φ<sub>f</sub> = weight of the flow

# Worst Case Weighted Fair Queueing (WF<sup>2</sup>Q)

• WFQ might get ahead of GPS



Worst Case Weighted Fair Queueing (WF<sup>2</sup>Q)



http://www.eng.tau.ac.il/~boaz/comnet/lec08.pdf

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# Worst Case Weighted Fair Queueing (WF<sup>2</sup>Q)



#### Worst Case Weighted Fair Queueing (WF<sup>2</sup>Q)



http://www.eng.tau.ac.il/~boaz/comnet/lec08.pdf

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#### Two-Dimensional Fair Queueing (known costs)

$W_1 \begin{bmatrix} a_1 & b_1 \\ W_0 \end{bmatrix}$	a 2 c 1	<i>b</i> <sub>2</sub>	<i>a</i> 3	b3	a 4 1 1	b,	a 5	bs c	a 6 2	<i>b</i> 6	a 7	b,	a <sub>8</sub>	b <sub>s</sub>	a9 c	
(a) Ideal request schedule over time on two threads																
Request Start Time Finish Time	<i>a</i> <sub>1</sub> 0 1	a2 1 2	<i>a</i> 3 2 3	a4 3 4	<i>a</i> 5	a <sub>6</sub> 5 6	a7 6 7	a <sub>8</sub> 7 8	<i>a</i> 9 8 9		S' Fir	Req tart 1 hish 1	uest Time Time	0 4	$c_2 c$ 4 4 8 1	3 3 2
Request Start Time Finish Time	<i>b</i> <sub>1</sub> 0 1	b2 1 2	<i>b</i> <sub>3</sub> 2 3	<i>b</i> <sub>4</sub> 3 4	<i>b</i> <sub>3</sub> 4 5	<i>b</i> <sub>6</sub> 5 6	<i>b</i> <sub>2</sub> 6 7	<i>b</i> <sub>8</sub> 7 8	<i>b</i> <sub>9</sub> 8 9		Request d <sub>1</sub> Start Time 0 Finish Time 4				$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	
$W_1 \begin{array}{c} b_1 \\ b_2 \\ W_0 \\ a_1 \\ a_1 \\ a_2 \\ a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_1 \\ a_5 \\ a_1 \\ a_2 \\ a_1 \\ a_2 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_1 \\ a_2 \\ a_2 \\ a_1 \\ a_2 \\ a_2 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_5 \\ a_1 \\ a_2 \\ a_5 \\ a_1 \\ a_2 \\ a_2 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a$	b) Request s $\frac{b_3}{b_4}$			$\frac{d_i}{c_i}$			$b_s$ $a_s$	$b_6$	b <sub>7</sub>	$b_8$ $a_8$	$d_2$			b, a,		
		(c) R	lequ	est	sche	dule	e pro	oduc	ced	und	er V	VFC	2			
$ \begin{array}{ccc} W_1 & b_1 \\ W_0 & a_1 \end{array} $	<i>d</i> <sub>1</sub> <i>c</i> <sub>1</sub>			$b_2$ $a_2$	b 3 a 3	b₄ a₄	b 5 a 5	$\begin{array}{c c} b_{5} & d_{2} \\ a_{5} & c_{2} \end{array}$				b 6 a 6	b 7 a 7	$b_s$ $a_8$	b9 a9	
	(	<b>d</b> ) R	equ	est s	che	dule	pro	duc	ed u	inde	er W	$/F^2$	2			

## Two-Dimensional Fair Queueing (known costs)

- WF<sup>2</sup>Q allows burstiness when multiple worker threads are free and only large requests are available.
- Small requests eligible for either all threads or no threads.
- New eligibility condition on thread i:  $S(r_i) (i / n)^* l_i$  where  $0 \le i < n$ .
- Small requests become eligible on high-index threads first and tend to be serviced before low-indexed threads can service them.



# Extended Two-Dimensional Fair Queuing (2DFQ<sup>E</sup>)

Pessimistic Cost Estimation

- Safer to estimate a cheap request as inexpensive than an expensive request as cheap
- Treat unpredictable tenants as expensive
- L<sup>i</sup><sub>max</sub> = cost of largest request
- c<sub>r</sub> = true cost of just-completed request
- If  $c_r > L^i_{max}$ , set  $L^i_{max} = c_r$
- Otherwise, set  $L_{max}^{i} = \alpha L_{max}^{i}$ , where  $\alpha < 1$  (but close to 1)

# Extended Two-Dimensional Fair Queuing (2DFQ<sup>E</sup>)

Bookkeeping: Retroactive Charging

- I<sub>r</sub> is the estimated cost of a request
- c<sub>r</sub> is the actual cost of a request
- Upon completion of a request, c<sub>r</sub> I<sub>r</sub> is incorporated into the virtual start and finish times of a tenant

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• This ensures long-run fairness

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# Extended Two-Dimensional Fair Queuing (2DFQ<sup>E</sup>)

Bookkeeping: Refresh Charging

- If a tenant transitions from many small requests to many expensive requests, L<sup>i</sup><sub>max</sub> will be underestimated at first
- If a request is especially expensive, up to N worker threads could begin processing expensive requests before the value of L<sup>i</sup><sub>max</sub> is updated
- Refresh Charging periodically checks long-running requests and updates L<sup>i</sup><sub>max</sub> while expensive request is still being processed
- Non-negligible overhead optimal at 10ms

# Evaluation of 2DFQ

- Discrete event simulator with synthetic workloads and traces from Azure Storage
- Comparing 2DFQ, WFQ and WF<sup>2</sup>Q
  - $\circ~$  SFQ, MSF², and DRR were also used in evaluation but omitted because their improvements minimally influence fairness bounds

<sup>(</sup>With slides adapted from author's SIGCOMM talk:

# **Evaluation Metrics**

Service lag: difference between service a tenant should have received with GPU (Nr, where N = number of threads and r = processing rate) and actual work done

Service rate: work done per 100ms

Latency: time between request being enqueued and finishing processing

Gini index: instantaneous measure of scheduling fairness

#### Evaluation with known cost



Synthetic data Costs known 16 threads 1000 units/second



Synthetic data Costs known 16 threads 1000 units/second 21







Synthetic data Costs known 16 threads 1000 units/second

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# Evaluation with known cost

- Using Azure production traces
  - 250 randomly chosen tenants from 50 servers, plus T<sub>1</sub>...T<sub>12</sub>
- Evaluate the effects on T<sub>1</sub>, a tenant with low request costs and low unpredictability



### Evaluation with known cost



# Evaluation with known cost



# Evaluation with known cost



# Evaluation with known cost



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### Evaluation with known cost



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#### Evaluation with unknown costs

- Using 2DFQ<sup>E</sup> with  $\alpha = 0.99$
- Added refresh and retroactive bookkeeping to WFQ and WF<sup>2</sup>Q to create WFQ<sup>E</sup> and WF<sup>2</sup>Q<sup>E</sup>

# Evaluation with unknown costs

- 300 randomly selected tenants from Azure data
- Added unpredictability by sampling from all Azure data without regard for server or account

#### 3 Experiments:

- 0% unpredictable: Only real tenant data
- 33% unpredictable: 33% arbitrary sampling
- 66% unpredictable: 66% arbitrary sampling

# Evaluation with unknown costs



# Evaluation with unknown costs



# Evaluation with unknown costs











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# Evaluation with unknown costs



Where  $t_1...t_7$  are fixed-cost tenants submitting requests of size  $2^8$ ,  $2^{10}$ ,  $2^{12}$ , ...,  $2^{20}$ 

#### Evaluation with unknown costs



# Evaluation with unknown cost: production workloads

- 150 experiments from Azure data
- Randomly vary several parameters:
  - Number of worker threads (2-64)
  - Number of tenants (0-400)
  - Replay speed (0.5-4x)
  - Number of backlogged tenants (0-100)
  - Number of artificially expensive tenants (0-100)
  - Number of unpredictable tenants (0-100)

# Evaluation with unknown cost: production workloads

99th percentile latency speedups

For T<sub>1</sub>, median improvement over WFQ<sup>E</sup> of 3.8x and 142x over WF<sup>2</sup>Q<sup>E</sup>

T<sub>10</sub> latencies were usually unimproved, but when they were it was by a large factor



0.1 1 10 100 2DFQ<sup>E</sup> 99% Latency (s) 46

### Evaluation with unknown cost: production workloads



Where  $t_1...t_7$  are fixed-cost tenants submitting requests of size 2<sup>8</sup>, 2<sup>10</sup>, 2<sup>12</sup>, ..., 2<sup>20</sup>

# Summary

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# Discussion

- Analyze the tradeoffs between how aggressively tenants with unpredictable costs and separated from predictable costs
- Can take a very long time to classify a previously expensive thread as a cheap one
  - Try changing alpha parameter to be dynamic
- Benefits of keeping system work-conserving
  - Always keep a set of threads dedicated to only serving inexpensive, predictable requests