

# MODEL ENSEMBLE TOOLS FOR SELF-MANAGEMENT IN DATA CENTERS

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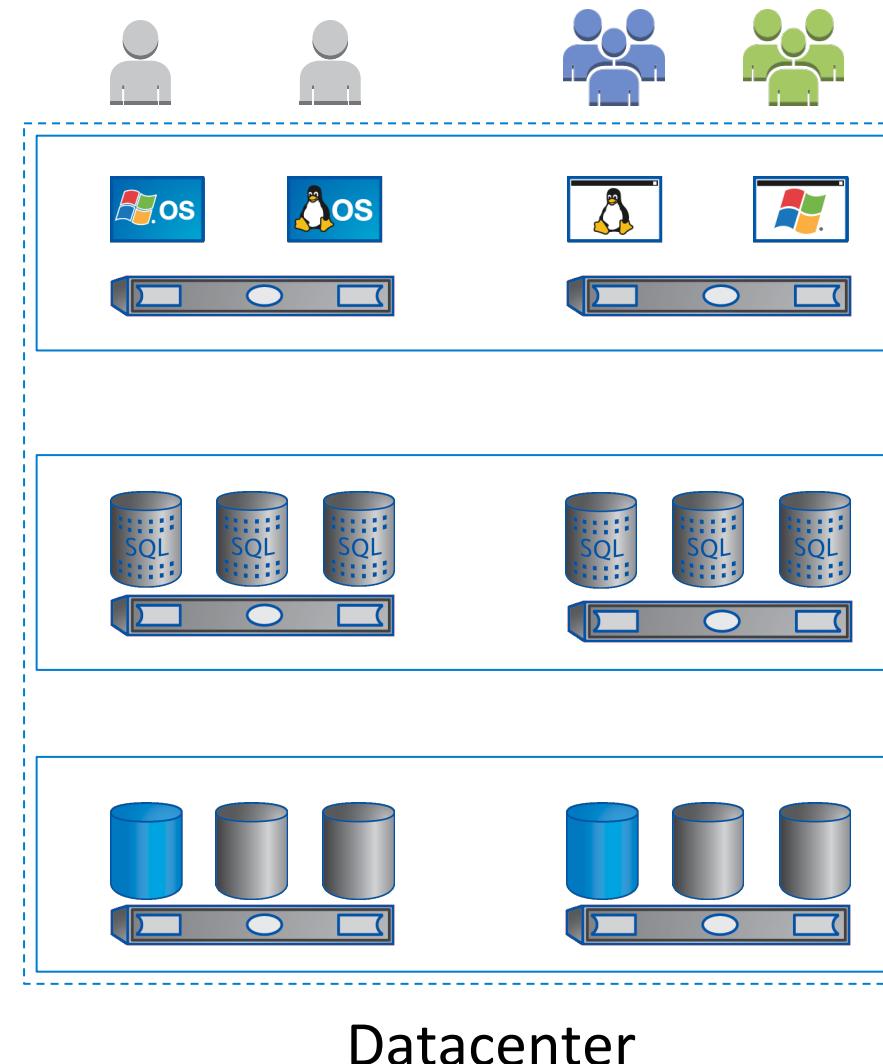
# Shared datacenter platforms



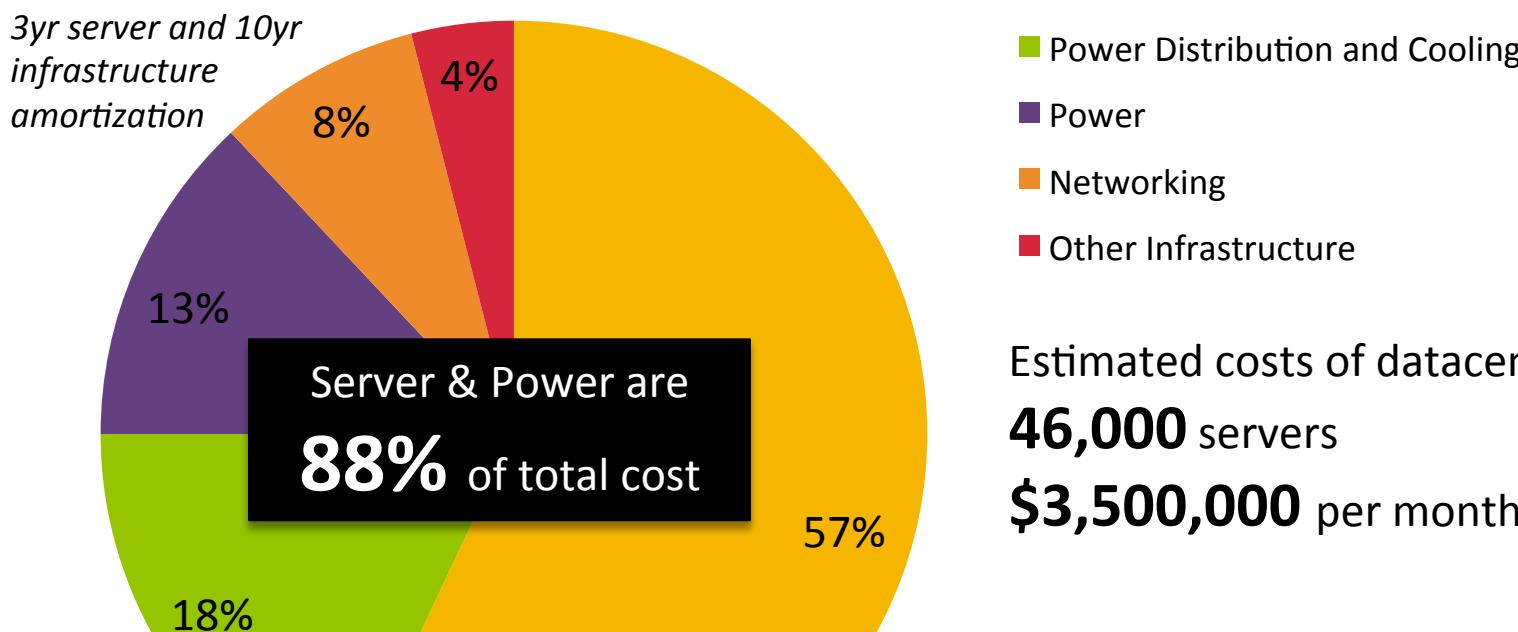
Amazon Web Services  
(Relational Data Service)



Google Compute Engine  
(Cloud SQL)



# Costs in a datacenter



- Server
- Power Distribution and Cooling
- Power
- Networking
- Other Infrastructure

Estimated costs of datacenter:  
**46,000** servers  
**\$3,500,000** per month to run

*Data courtesy of James Hamilton [SIGMOD'11 Keynote]*

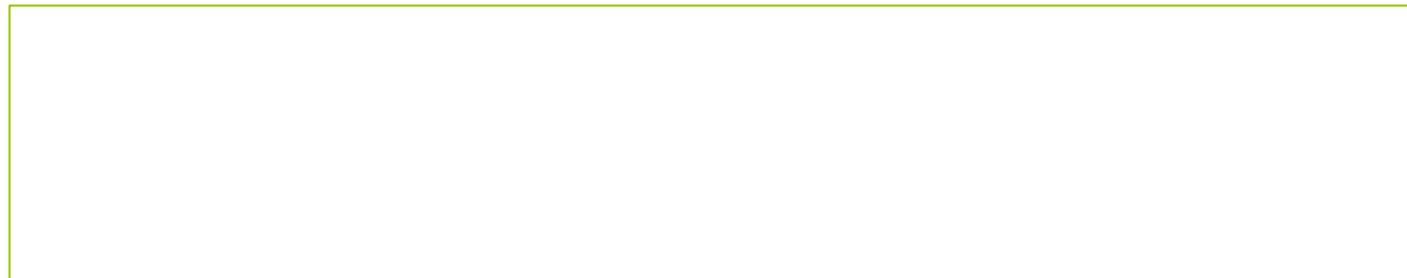
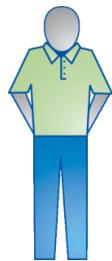
# Ways to reduce costs

Service Provider Administrator

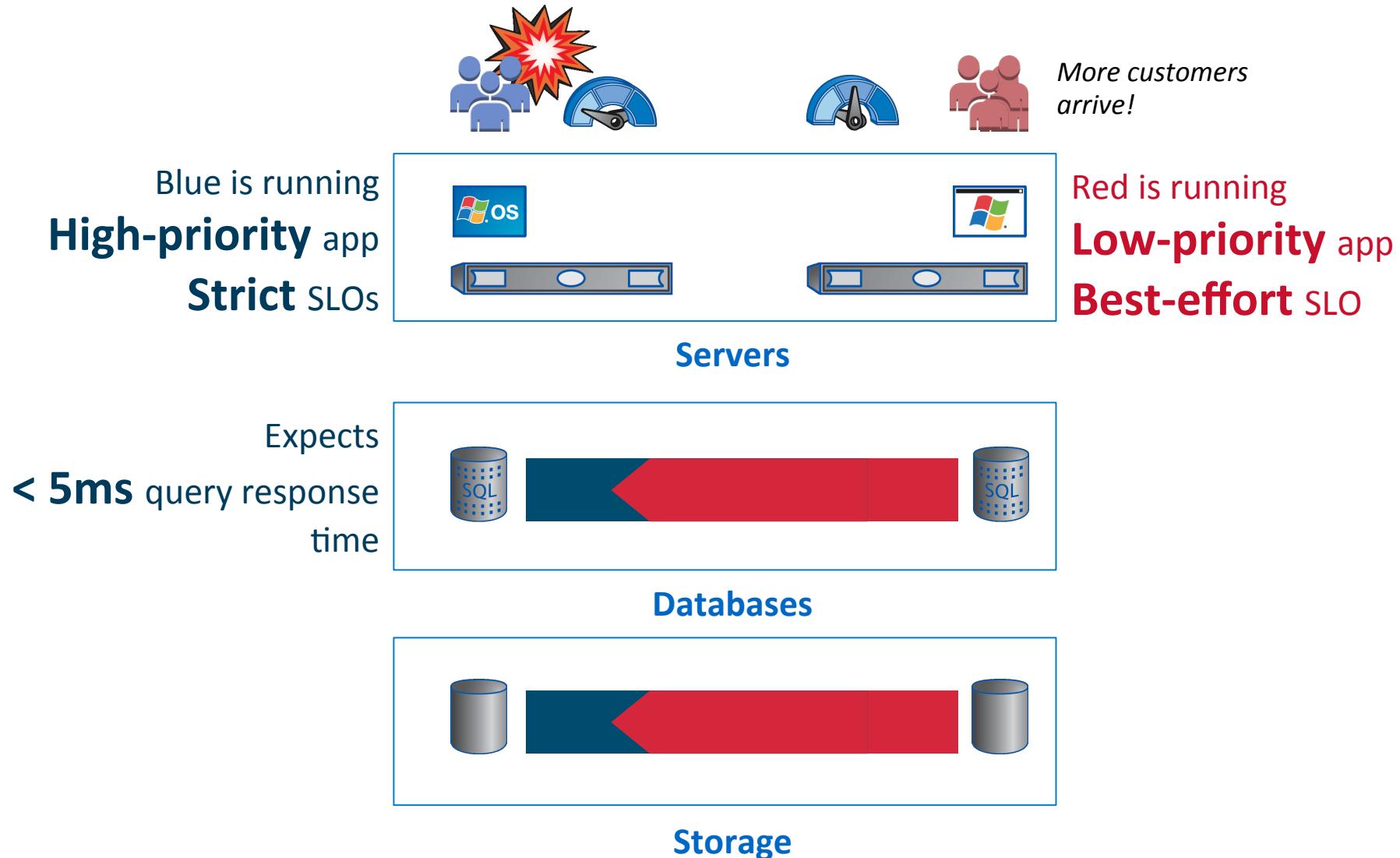


- **Use less resources:** Customer wants 1000 TPS. What is the most efficient (e.g., CPU/Memory) to deliver it?
- **Share resources:** Can I place customer A's DB along side customer B's DB? Will their service-levels be met?

Customer DBA



# Provider: Support multiple customers



# How to reduce costs

Service Provider Administrator



- **Use less resources:** Customer wants 1000 TPS. What is the most efficient (e.g., CPU/Memory) to deliver it?
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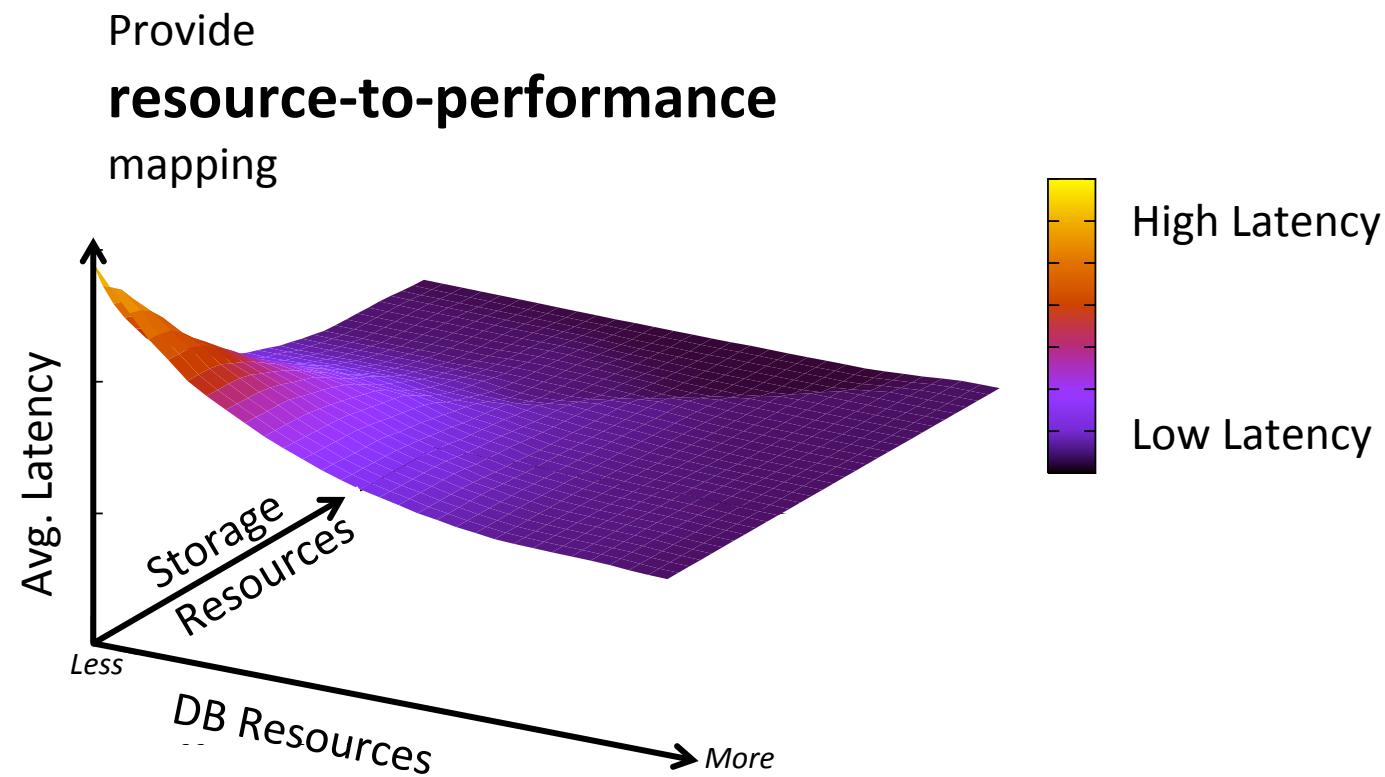
Customer DBA



- **Use the right amount of resources:** What will be the performance (e.g., query latency) if I use 8GB of RAM instead of 16GB?
- **Solve performance problems:** I'm only getting 500 TPS. What's wrong? Is the cloud to blame?

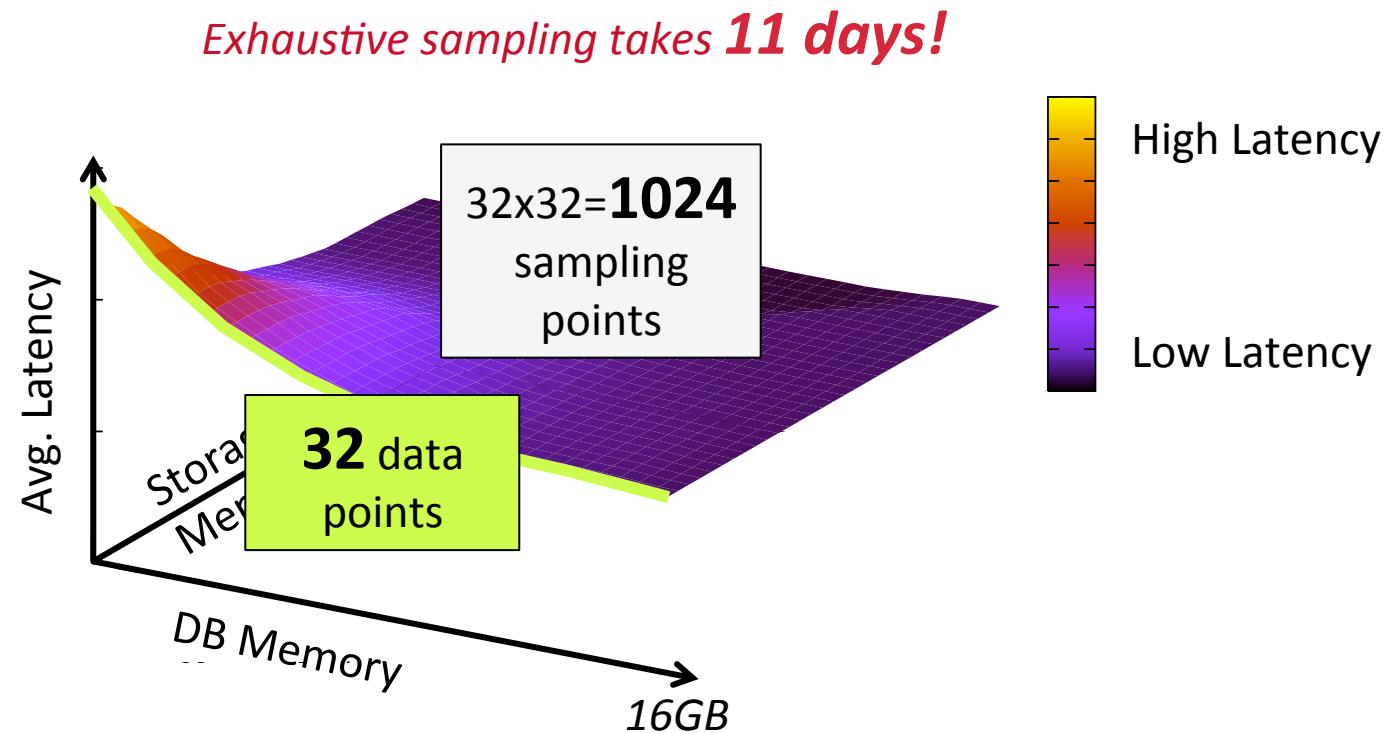
Need to build **performance models** to understand

# What are performance models?



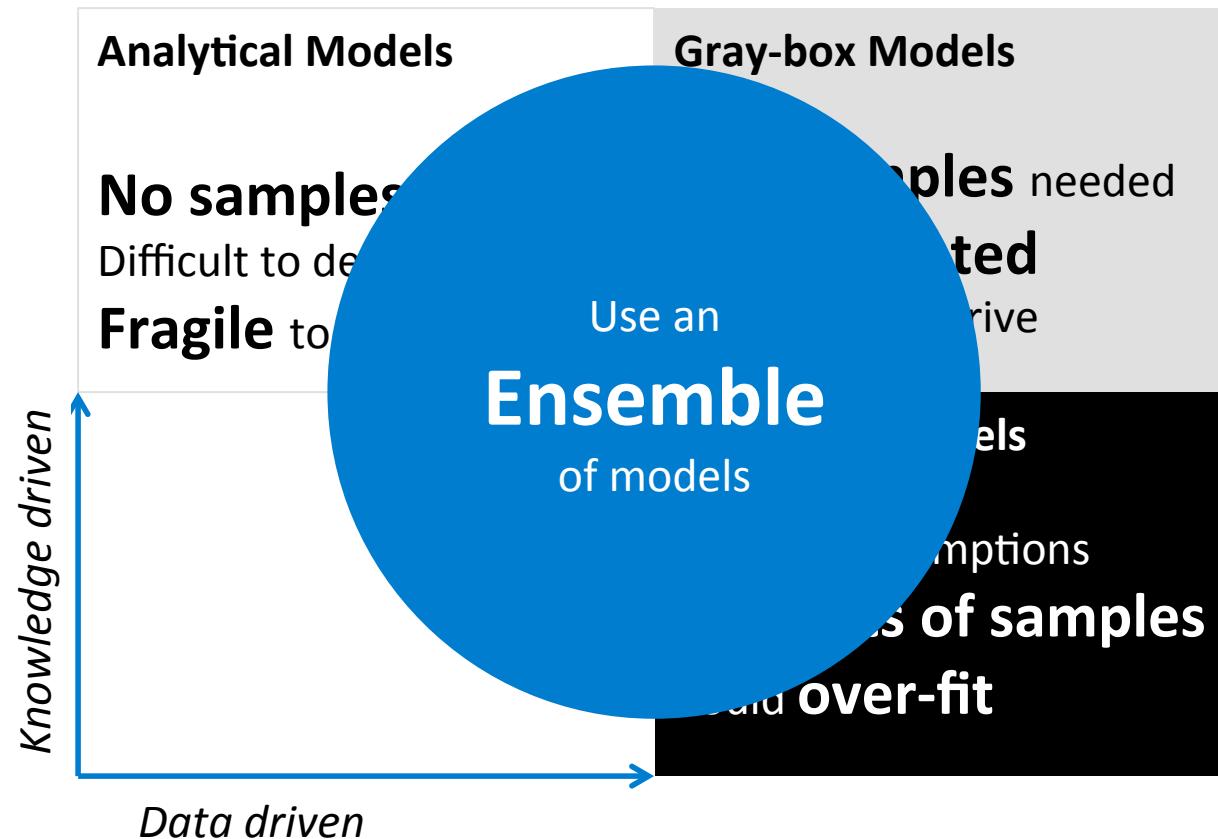
- ✓ Dynamic Resource Allocation [FAST'09]
- ✓ Capacity Planning [SIGMOD'13]
- ✓ What-if Queries [SIGMETRICS'10]
- ✓ Performance Anomaly Detection [SIGMETRICS'10]

# Building performance models takes time



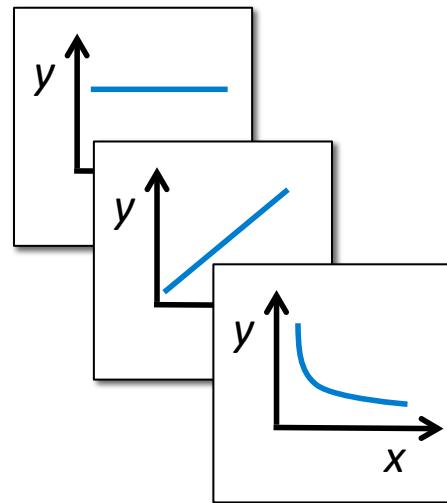
*Actuate a live system and take experimental samples.  
Sample in 512MB chunks; 15 minutes for each point*

# Types of performance models



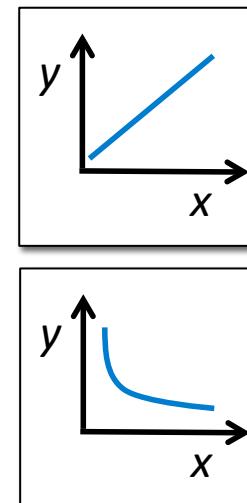
# Ensemble approach

**1.** Specify **trends** and **patterns**



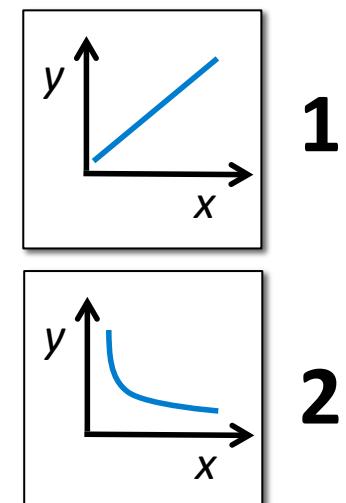
**2.** Automatically tune the models

*Use data*



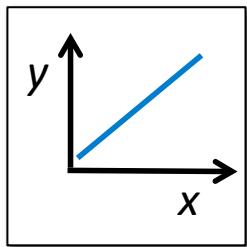
**3.** Rank the models use a **blend**

*Test & Rank*



*Repeat  
(if needed)*

# Specifying trends and patterns



$$\hat{y} = \alpha x_i + \beta$$

## SelfTalk

Language to **describe**  
**relationships**

Provide a **catalog** of  
**common functions**

```

DEFINE RELATION linear {
  PARAMETER a,b : number,
  INPUT x:number-array, y:number-array,
  ...
  FUNCTION confidence
  {
    OUTPUT confidence:number
    LANGUAGE 'matlab'
    SCRIPT
    $                               Goodness-of-fit
      y_hat = a.*x .+ b;
      confidence = R2(y,y_hat);
      //calculate residuals
      ...
    $
  }
  ...
}

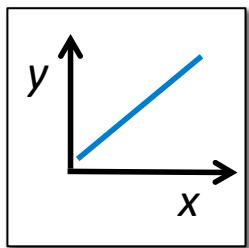
```

Specifies model  
**inputs** and  
**parameters**

**Curve-fitting**  
and validation  
**algorithms**

*Details in SIGMETRICS'10 paper*

# Refine models using data



$$\hat{y} = \alpha x_i + \beta$$

Use **hints** to **link** relations to metrics

```
HINT myHint
RELATION LINEAR(x,y)
METRIC (x,y) {
    x.name=MySQL.CPU
    y.name=MySQL.QPS
}
CONTEXT (a) {
    a.name=MySQL.BufPoolAlloc
    a.value >= 512MB
}
```

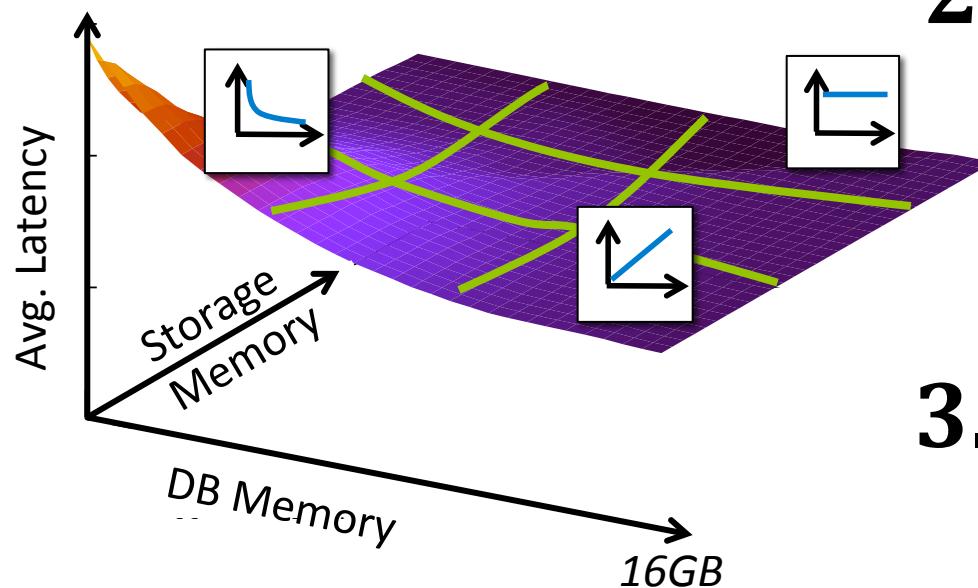
**Hints** that CPU  
linearly correlated  
to QPS

But **working-set**  
should be in **RAM**

**Learns parameters using data**  
(or requests more data)

# Rank models and blend

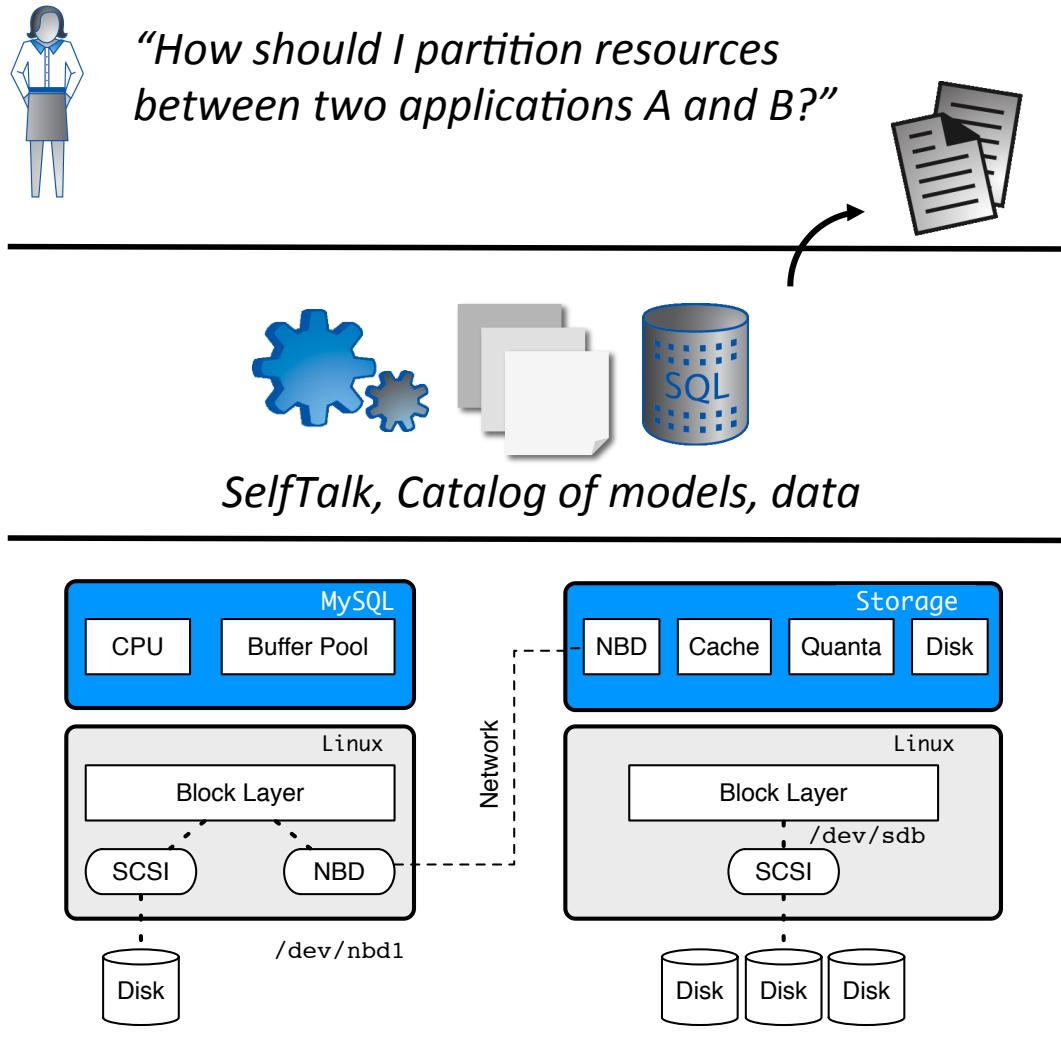
**1. Divide** search space  
into **regions**



**2. n-fold cross-validation**  
to **rank**

**3. Associate best-model to**  
**region**

# Prototype



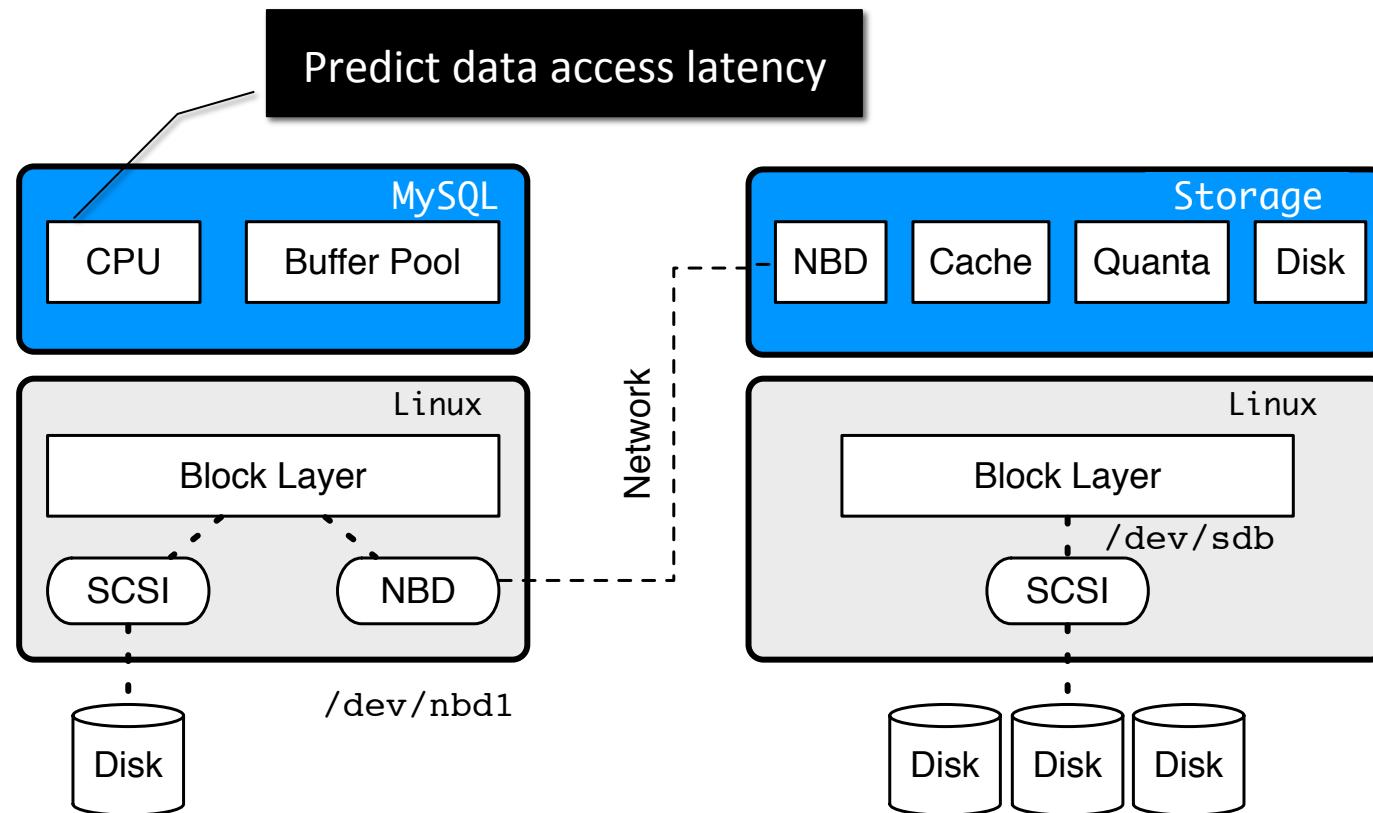
# Set of model templates

- Analytical models
  - A-STOR: analytical model for predicting storage hierarchy latency [FAST'09]
- Black-box models
  - B-SVM: support vector machine regression
  - B-CNST: average values as prediction results
- Gray-box models
  - G-RGN: region-based curve fitting model
  - G-INV: inversely-proportional curve fitting model

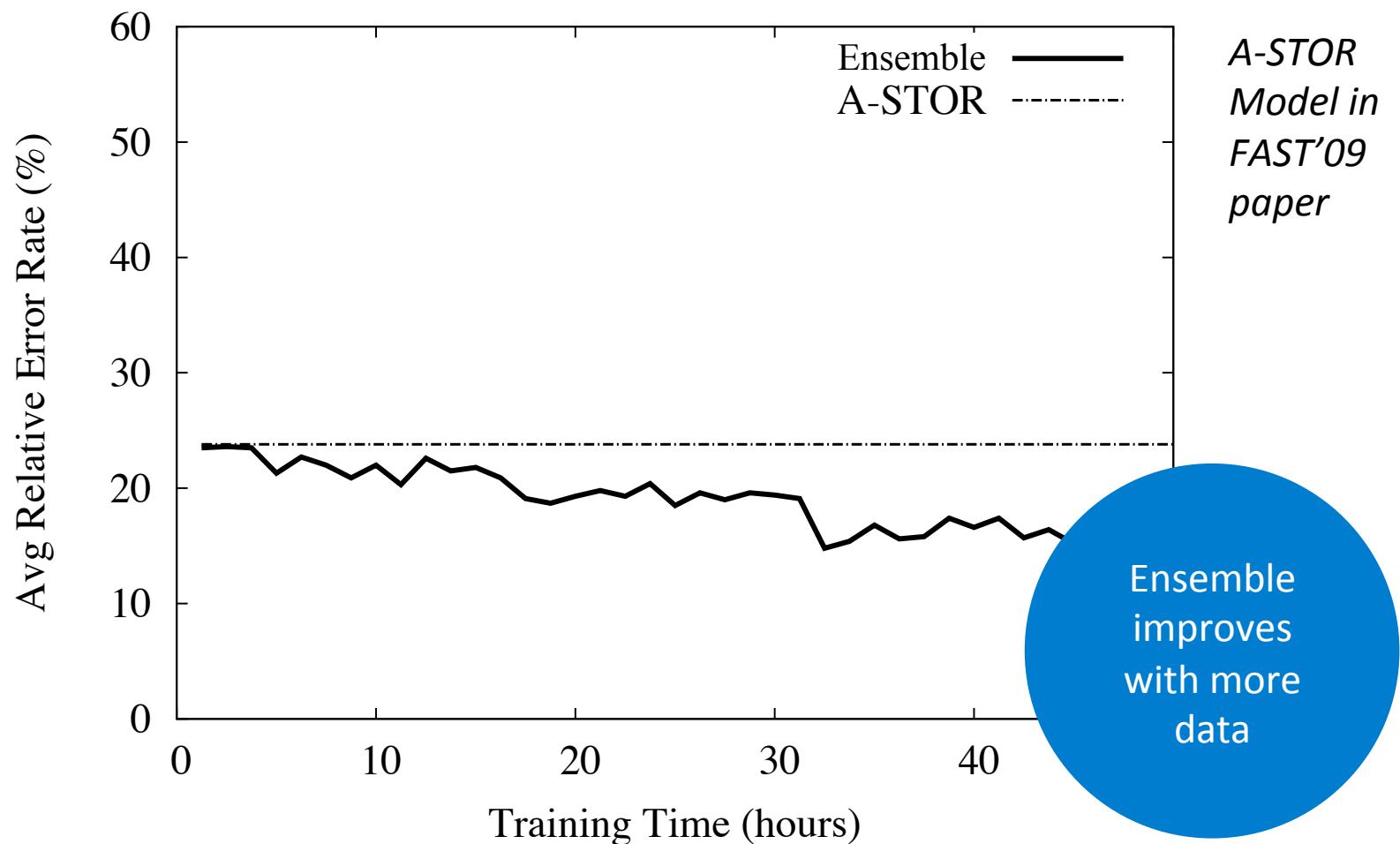
# Workloads and datasets

- Benchmarks
  - **TPC-W:** E-commerce (online bookstore) benchmark
  - **TPC-C:** Wholesale parts supplier with warehouses and sales districts
- Experiments
  - Conducted over a period of **6 months**
  - Data from OS, MySQL, and storage server
  - At the end, the structured data was **10GB**

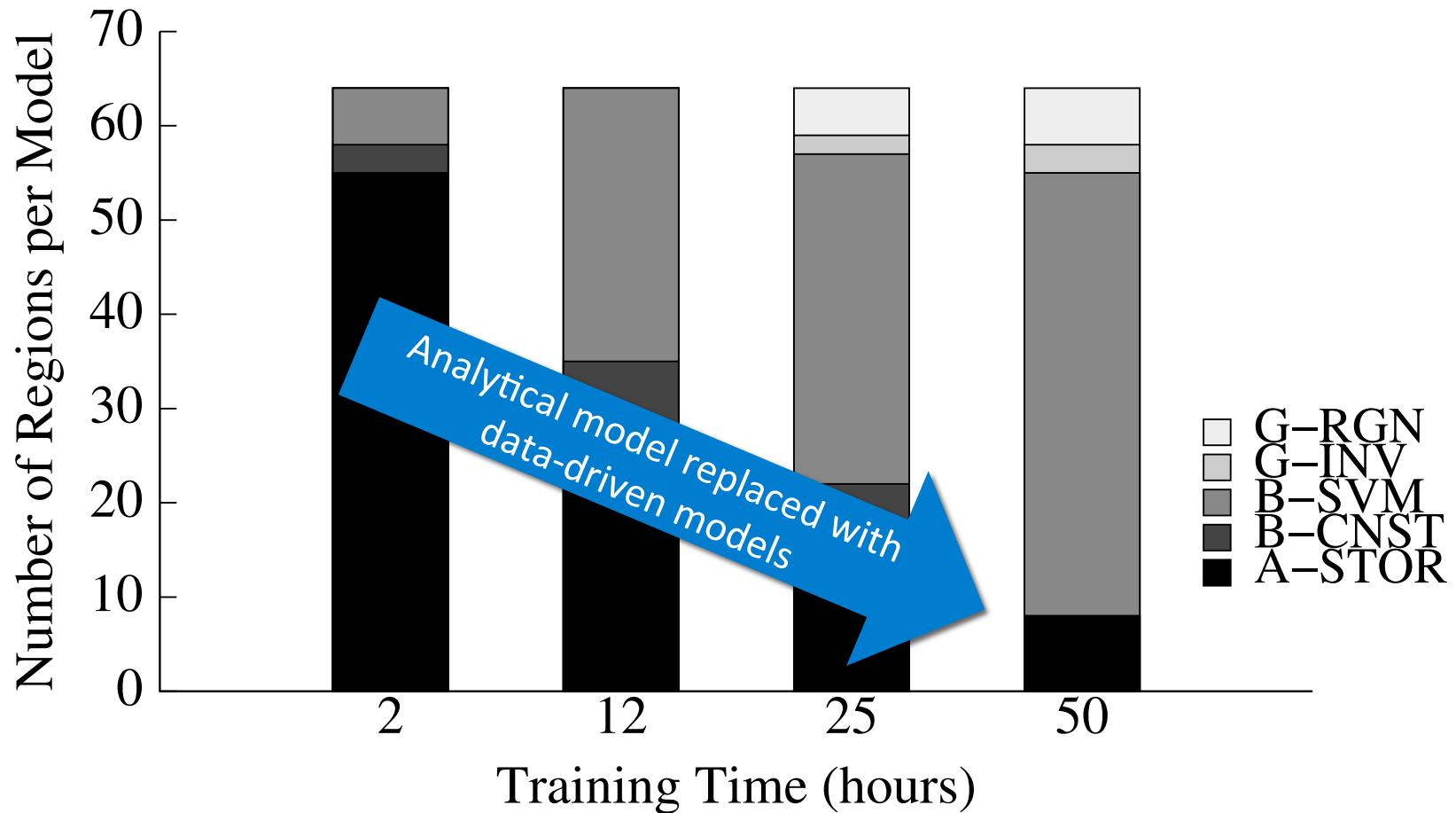
# Modeling MySQL IO subsystem



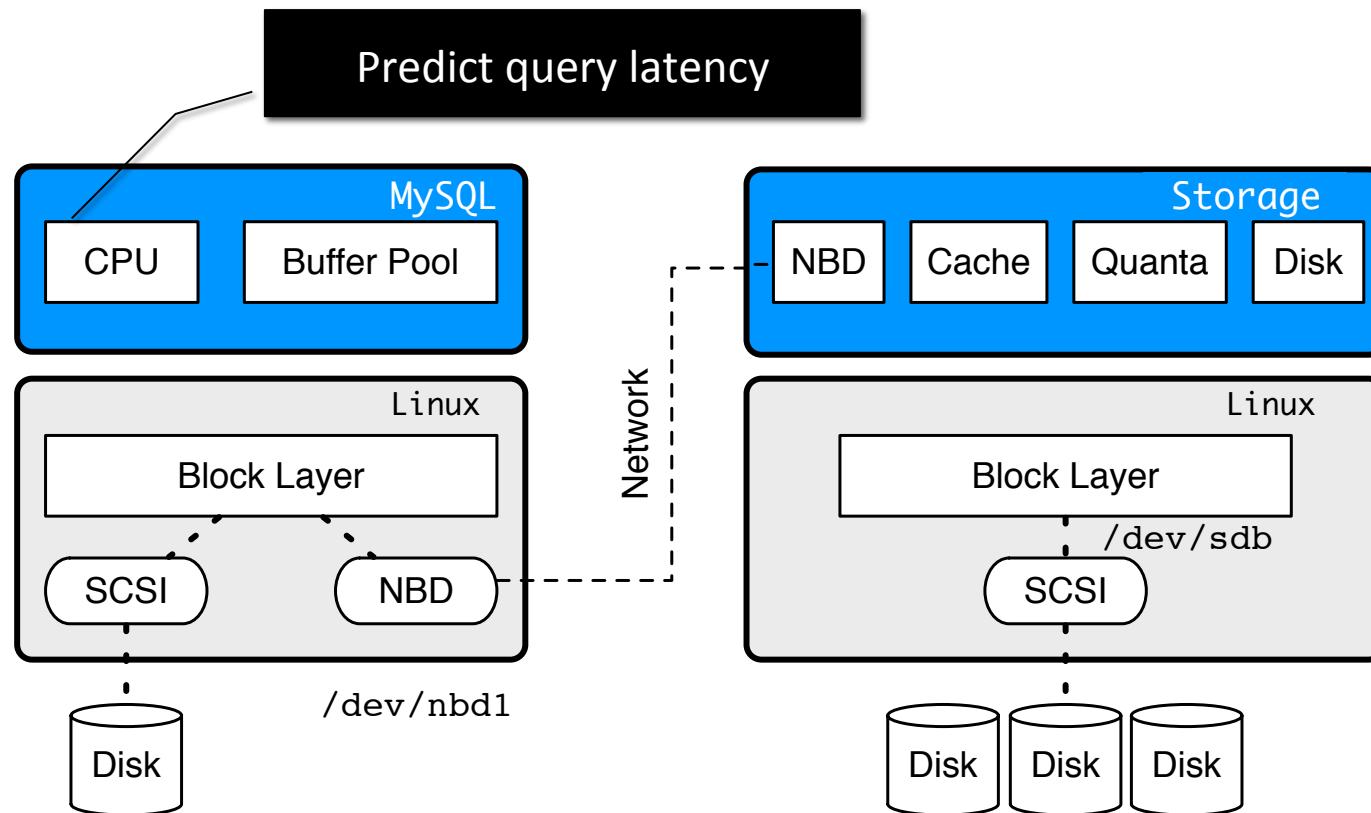
# Predicting TPC-W buffer pool latencies



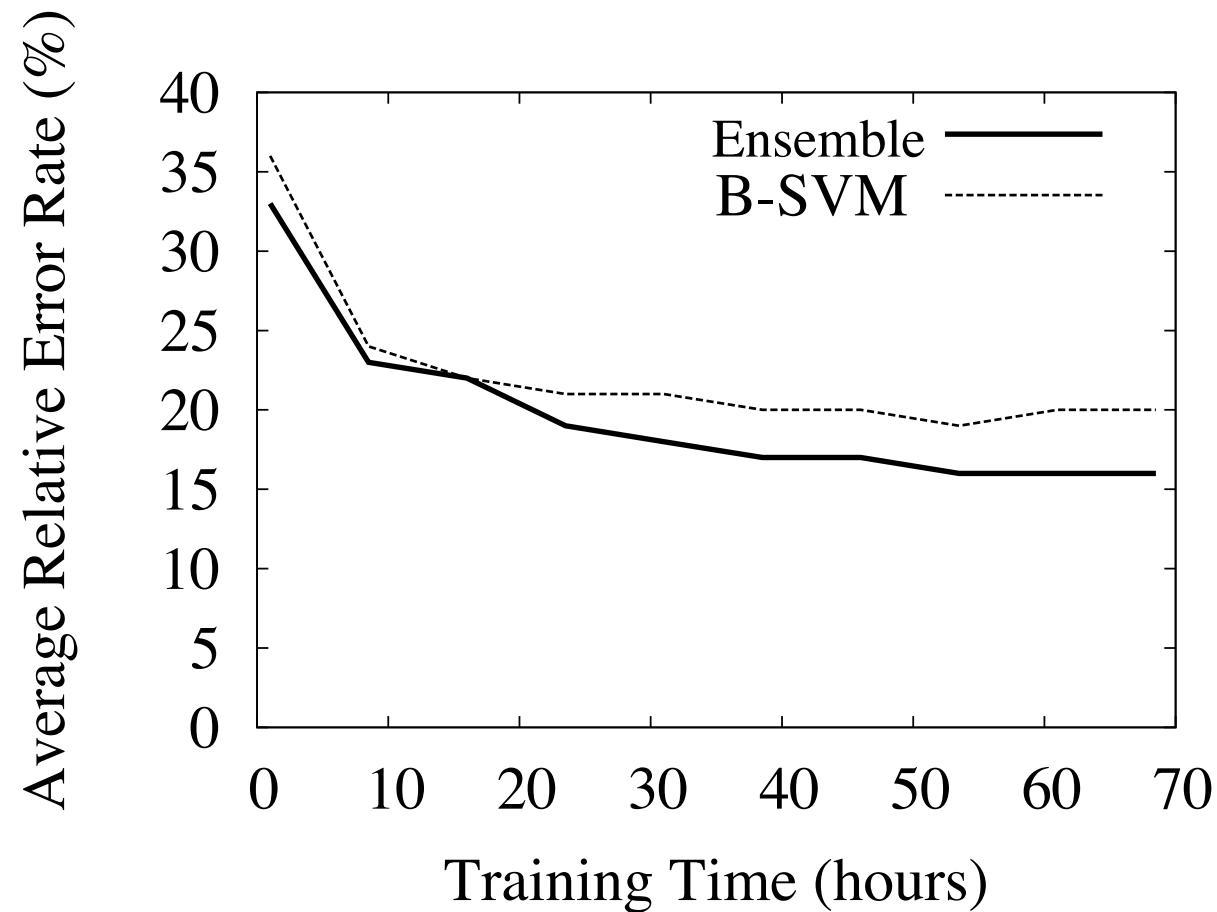
# Predicting TPC-W buffer pool latencies



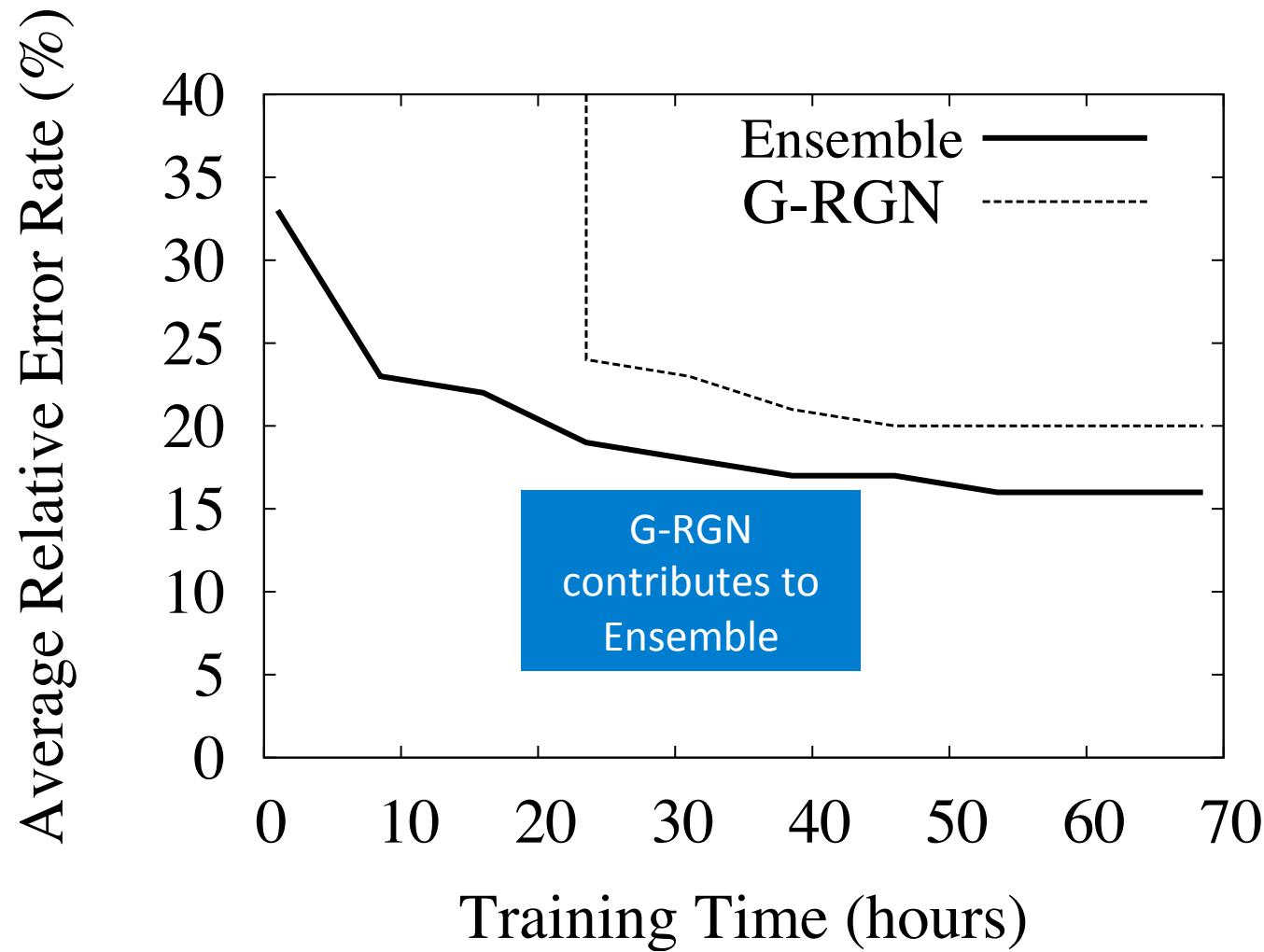
# Modeling MySQL end-to-end



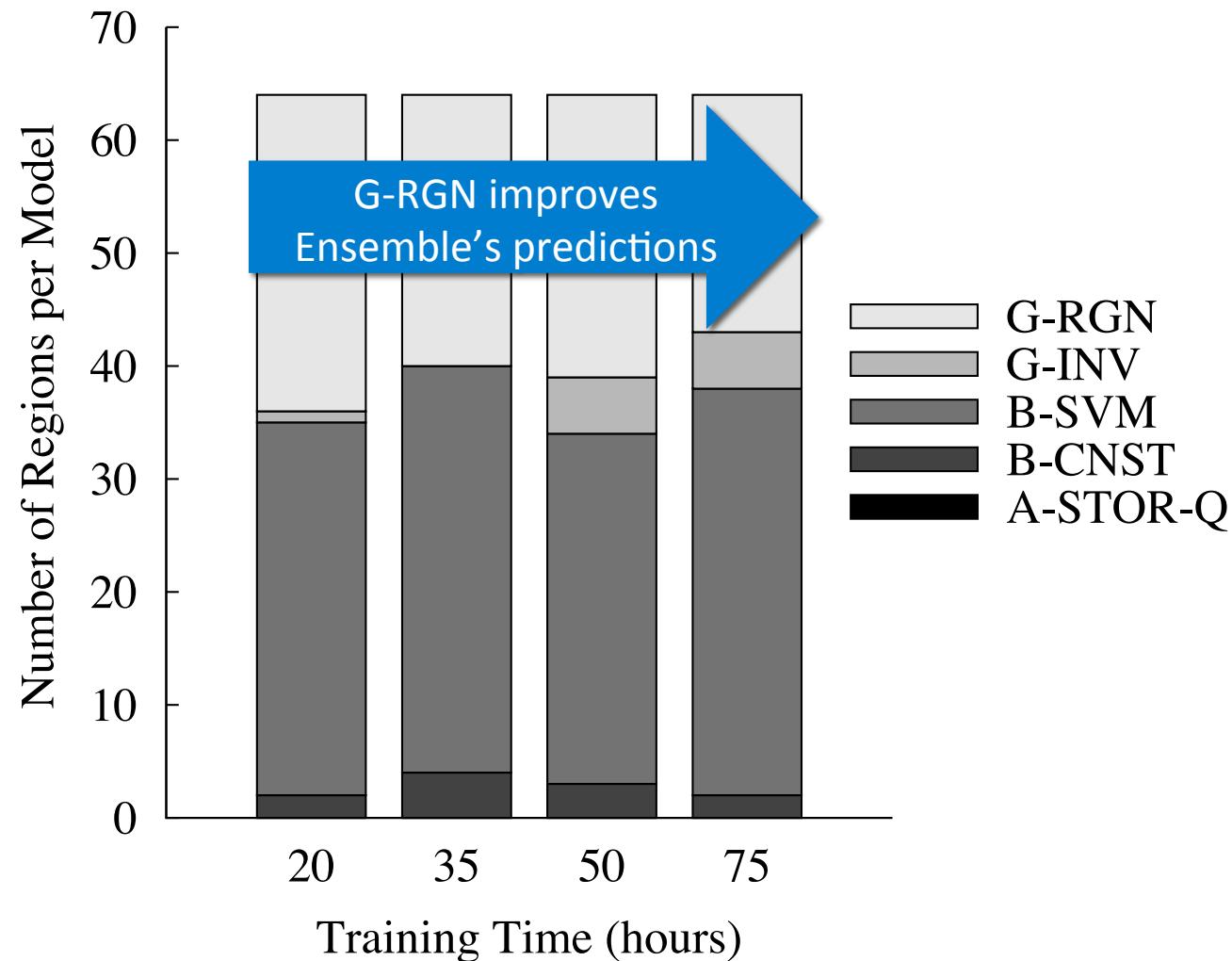
# Predicting TPC-C query latencies



# Predicting TPC-C query latencies



# Predicting TPC-C query latencies



# Conclusions

- We built Ensemble
  - An **interactive modeling** framework
  - Uses **semantic knowledge** as model guides
  - Runtime engine iteratively validates, selectively reuses and refines models
- Applied Ensemble to build performance models
  - Allows one to **explore tradeoffs** of analytical, black-box, and gray-box models
  - Showed that we can create models with better accuracy in shorter periods of time

# References

- “Ensembles of Models for Automated Diagnosis of System Performance Problems” DSN’05
- “IRONModel: Robust Performance Models in the Wild” SIGMETRICS’08
- “*Dynamic Partitioning of the Cache Hierarchy in Shared Data Centers*” VLDB’08
- “*Dynamic Resource Allocation for Database Servers Running on Virtual Storage*” FAST’09
- “*A Query Language and Runtime Tool for Evaluating Behavior of Multi-tier Servers*” SIGMETRICS’10
- “Performance and Resource Modeling in Highly-Concurrent OLTP Workloads” SIGMOD’13

Thank you!

# Mixing CPU and Disk models

