

# **Towards a Unified Architecture for in-RDBMS Analytics**

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CS 294: Big Data Systems Research

# **Motivation**

# Advanced Analytics

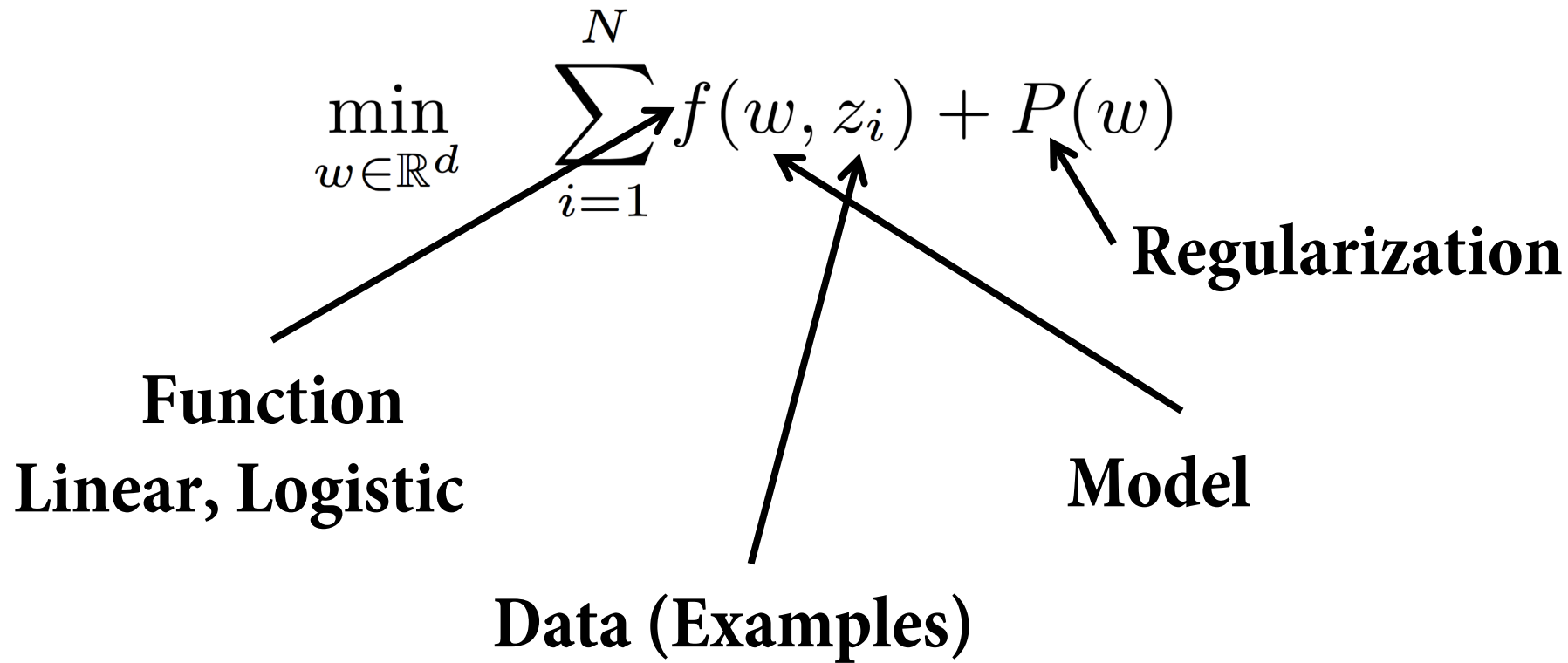
**Classification**



**Recommendation**



# Convex Optimization



# Convex Optimization

$$\min_{w \in \mathbb{R}^d} \sum_{i=1}^N f(w, z_i) + P(w)$$

**What is convex ?**

**Linear Regression, Linear SVM**

**Kernel SVMs, Logistic Regression,**

**What is not convex ?**

**Graph mining, Deep Learning**

# Gradient Descent

$$w^{(k+1)} = w^{(k)} - \alpha_k \nabla f(w^{(k)})$$

**Initialize w**

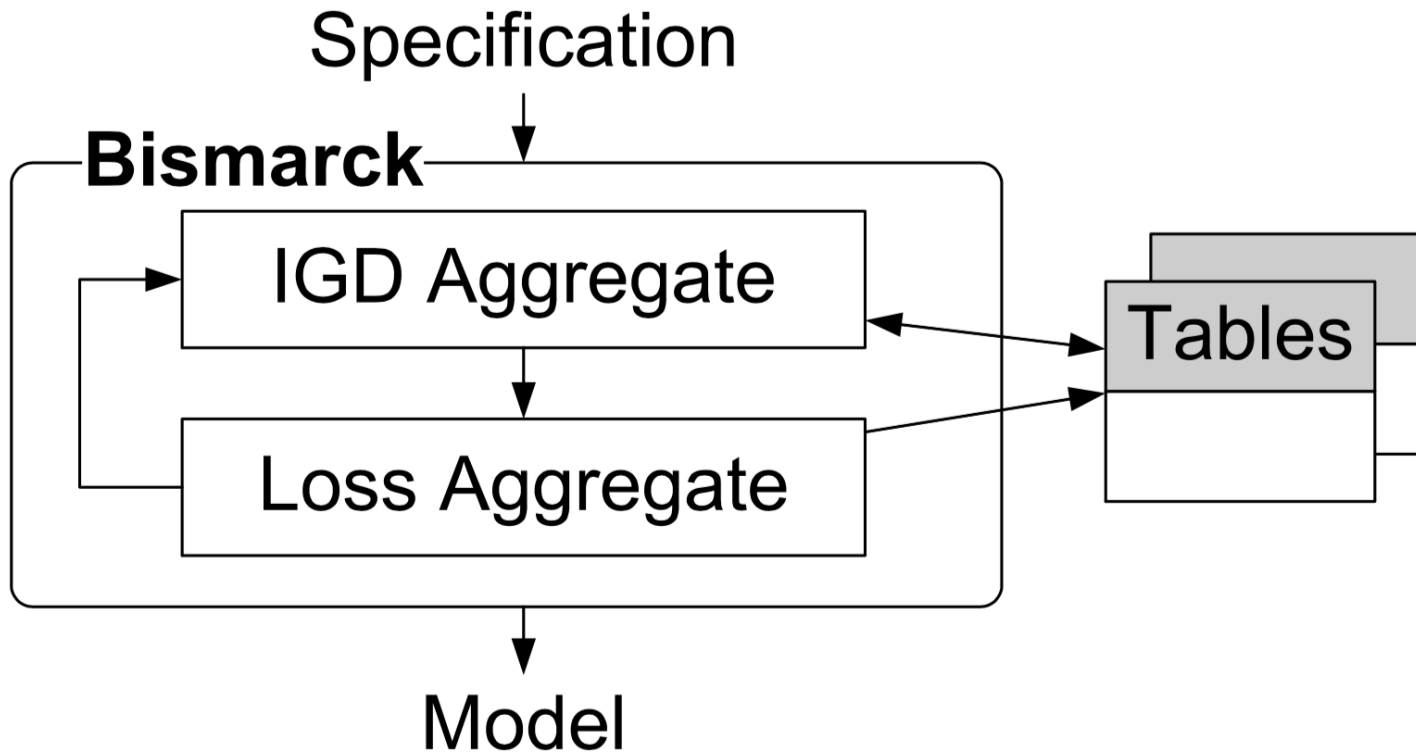
**For many iterations:**

**Compute Gradient**

**Update model**

**End**

# Bismarck Architecture



# Spark MLlib

## Gradient

(data, weights) -> gradient

## Updater

(weights, grad) -> newWeights

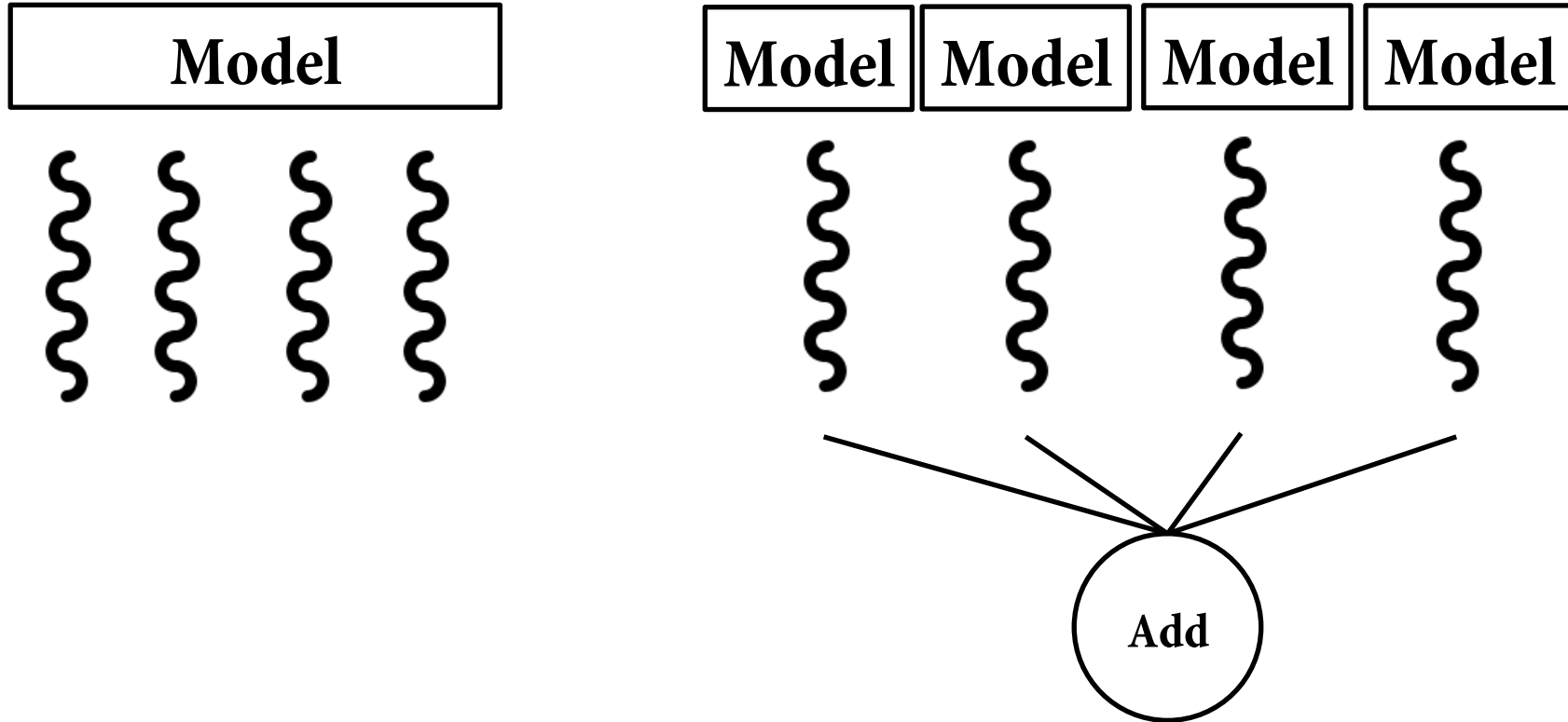
## Gradient Descent

```
for (i <- 1 to ITERS) {  
  val gradSum = data.sample(fraction).map { x =>  
    gradient.compute(x, weights)  
  }.reduce(_ + _)  
  
  weights = updater.compute(weights, gradSum)  
}
```



# Discussion

# (1) Shared Memory vs. Shared Nothing



**Bismarck: Shared memory beats Shared Nothing**

# **(1) Shared Memory vs. Shared Nothing**

**Better Algorithms**

**AdaGrad**

**Dual Averaging**

...

**Better Systems**

**Parameter Server**

**AllReduce (VW)**

...

**Fundamental trade-off between delay & convergence**

## (2) Single disk vs. Cluster



<b>Compute Bandwidth</b>	<b>4-8 GB/s ? (PCIe)</b>	<b>60 GB/s * N</b>
<b>Update Latency (1MB)</b>	<b>30 microseconds</b>	<b>800 microseconds * N (10 Gbps)</b>

## (2) Single disk vs. Cluster

**Example: 1M data points, 128k features, 1 class**

**Data size: 1TB**

**Cluster: 8 machines ~ 128GB / machine**

**Single disk: 8 SSDs**

	<b>Single Disk</b>	<b>Cluster</b>
<b>1 Epoch Gradient</b>	<b>128s</b>	<b>2.5s</b>
<b>Updates</b>	<b>30 microseconds</b>	<b>6.4 milliseconds</b>

## **(2) Single disk vs. Cluster**

**More gradient computations**

**vs.**

**Frequent Updates**

**Depends on**

**Cost of gradient computation**

**Model Size**

**Data characteristics (condition number)**

# **(3) IGM vs. L-BFGS vs. SDCA etc.**

**One architecture to rule them all ?**

**Momentum, L-BFGS – Need for model history**

**Block Coordinate – Residuals (data size)**

**Dual Coordinate Ascent – Store dual coefficients**

**...**

# **(4) Efficient Sampling**

**Single node**

**Shuffle once (Bismarck)**

**Shuffle every epoch**

**Cluster**

**Bernoulli Sample per-batch**

**Need to look at entire data**

**Shuffle + Sample blocks (KMN)**