

### Data Intelligence for All

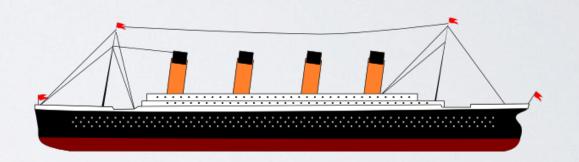
Christopher Nguyen, PhD Co-Founder & CEO

Presented on December 2, 2013

### What do you get when you cross ...







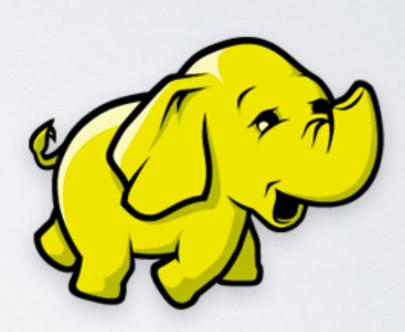
Titanic



### What do you get when you cross ...







## What do users need?

## Let's take a look!

... and that's what we're working on at Adatao

In the beginning, there was darkness...

### Then came Business Intelligence

### You could see what happened with your business



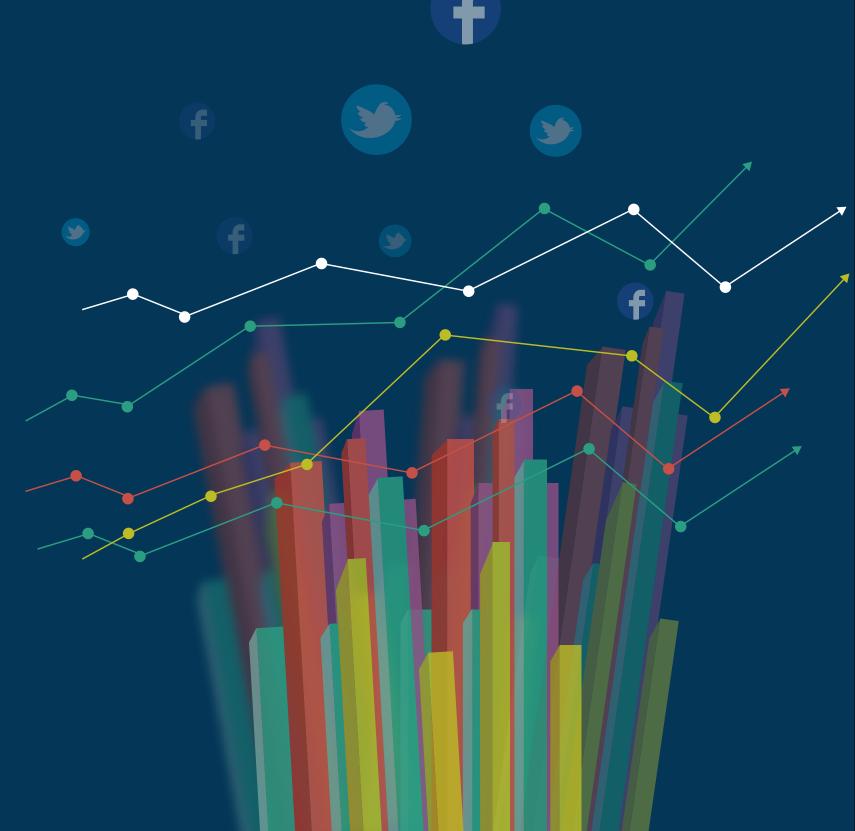


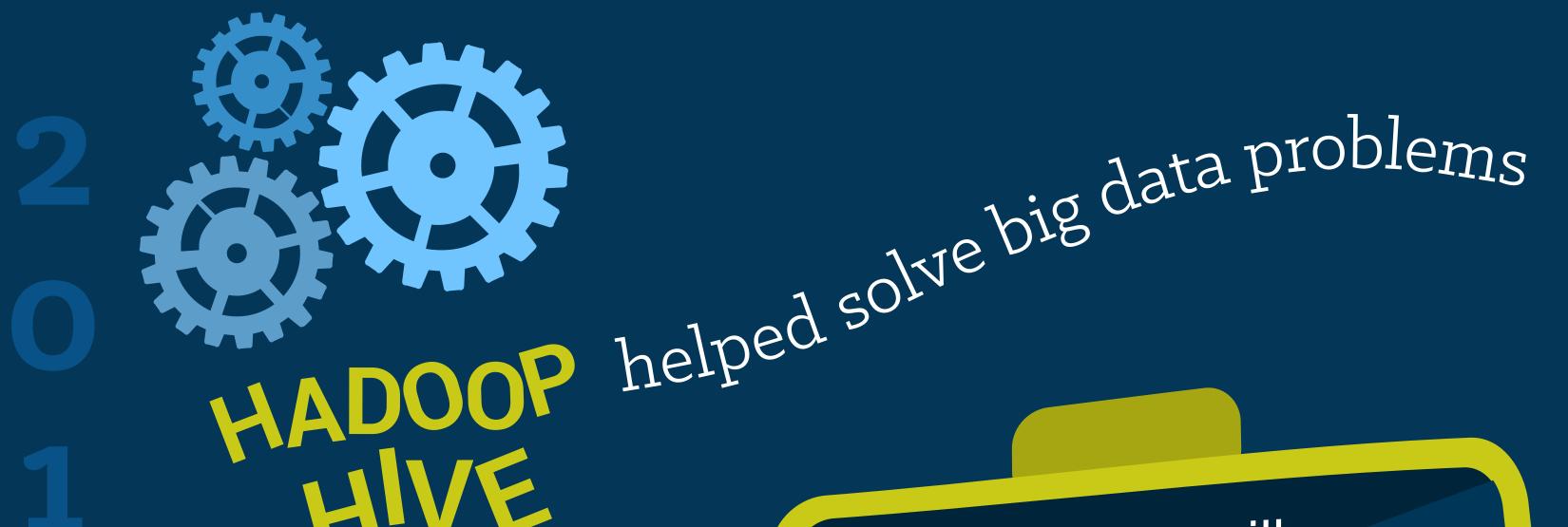
# BIG DATA PROBLEMS

Huge Volume

High Velocity

Great Variety



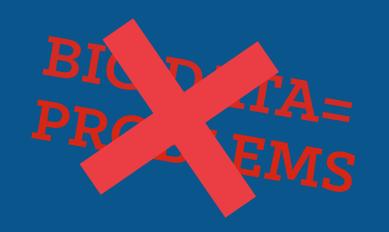


MAPREDUCE

but businesses still LOOKING BACKWARD



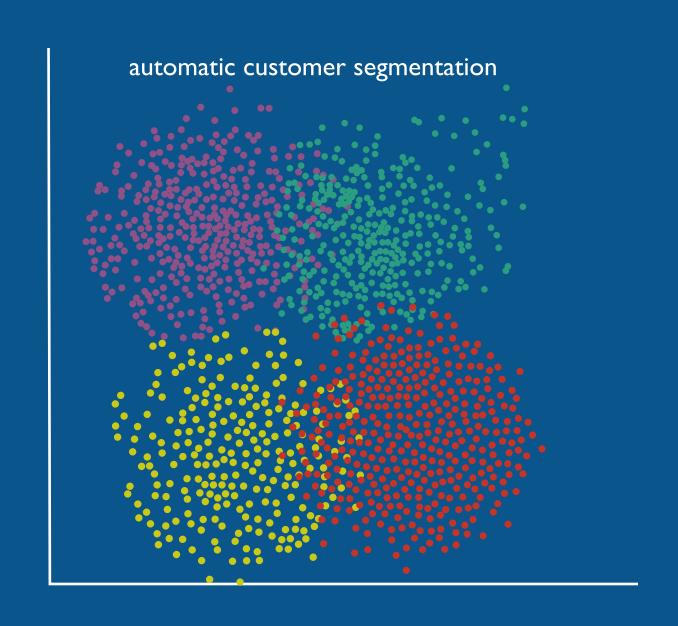
## BIG DATA + BIG COMPUTE = OPPORTUNITIES



Machine Learning

Predictive Analytics

Natural Language



#### **Business Users**

#### **Data Scientists**

#### **Data Engineers**



BIG INSIGHTS Visually Beautiful
Interactive Data
Exploration
Narrative Web App







BIG COMPUTE

Powerful In-Memory Data Mining
Machine Learning Big Analytics Platform



BIG DATA (Hadoop HDFS, Cassandra, SQL DMBS, Streaming Data)





Deep engineering & business experience from Google, Yahoo et al. PhD's in DM & ML from UIUC, Georgia Tech, Stanford, Berkeley, ...



Big-Data Compute Engines, Google Apps Engineering Director, Google Founders' Award, HKUST Prof, 2 successful enterprise exits, Stanford PhD



Hadoop distributed/streaming analytics, Yahoo Hadoop Eng, UIUC PhD

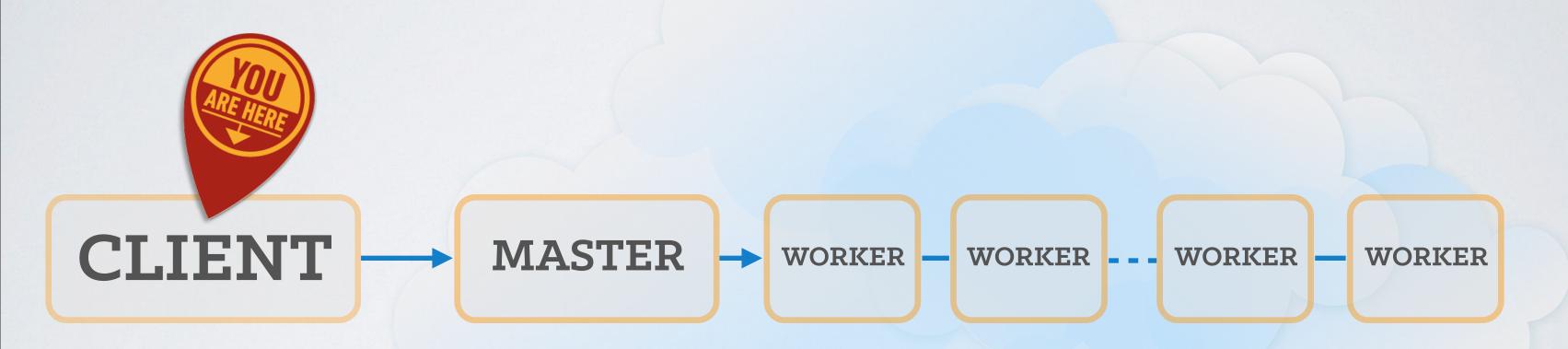


Machine learning & machine vision, US Army Research Lab, Johns Hopkins PhD

## Adatao pInsight demo



# Demo Deployment Diagram



## Adatao pAnalytics demo



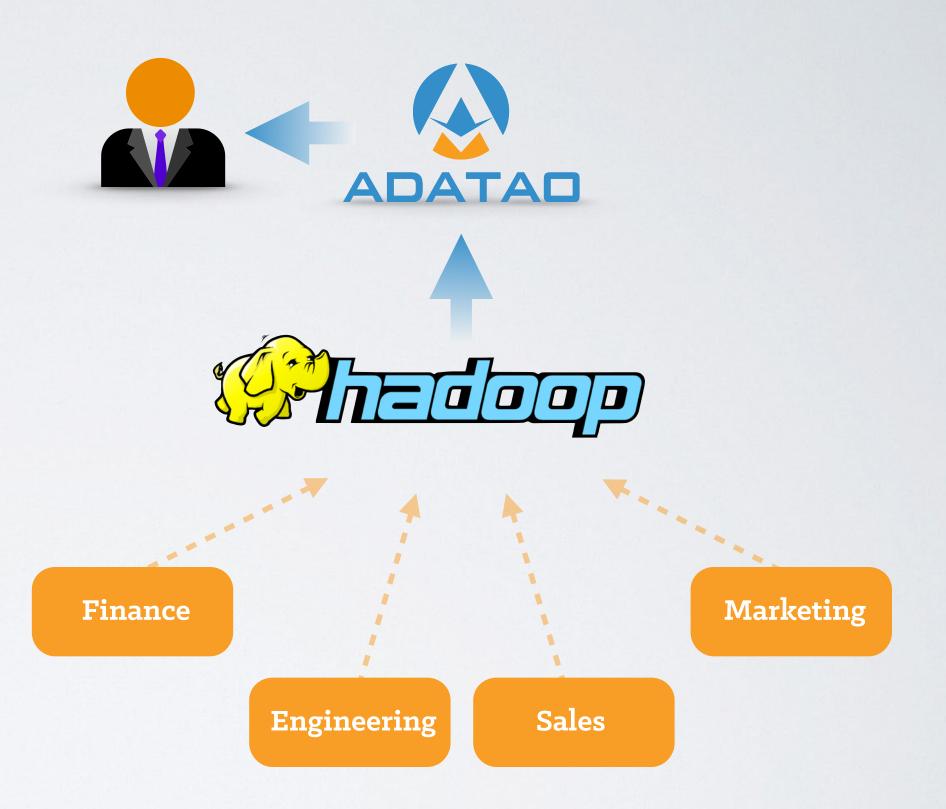
### Bonus Demo

### Use Cases

## Internet Service Provider

Interactive, Ad Hoc Business Query

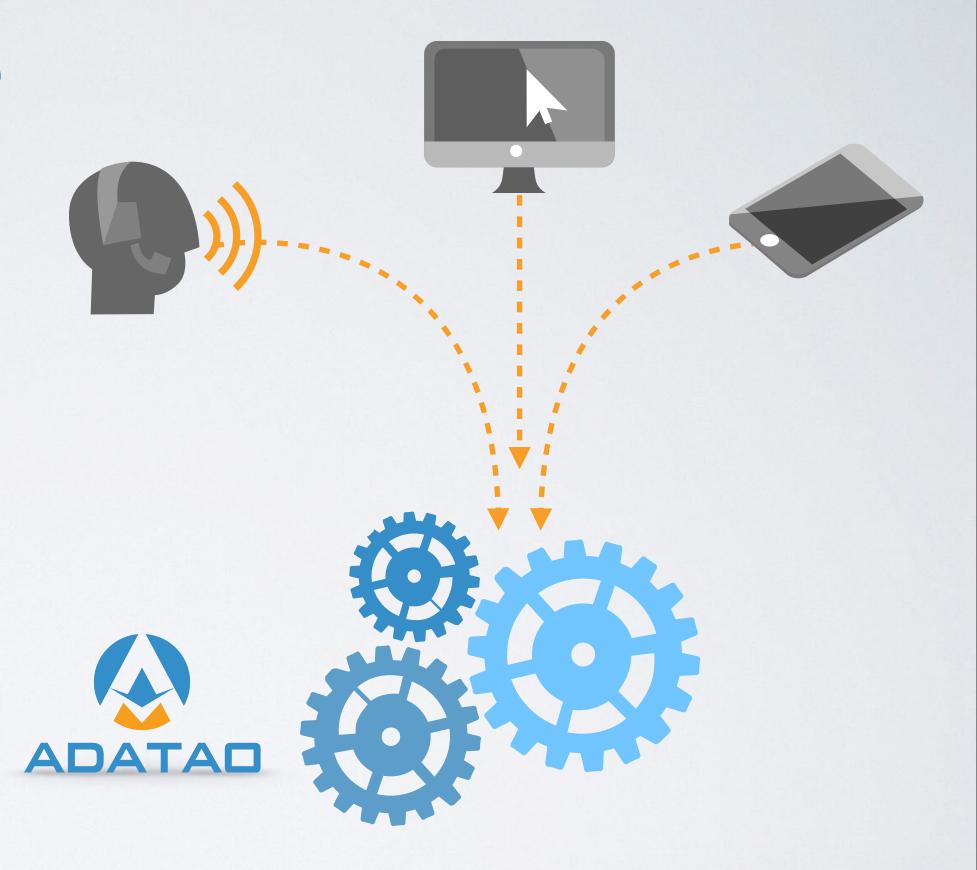
Insight Discovery on Aggregated Operational Data



## Customer Service Provider

Product Recommendation

Cross-channel
User Experience Optimization



## Heavy Equipment Manufacturer

Sensor Network Analytics for Predictive Maintenance



# Mobile Ad Platform

Ad Targeting

**CTR Prediction** 





## Scaling Performance

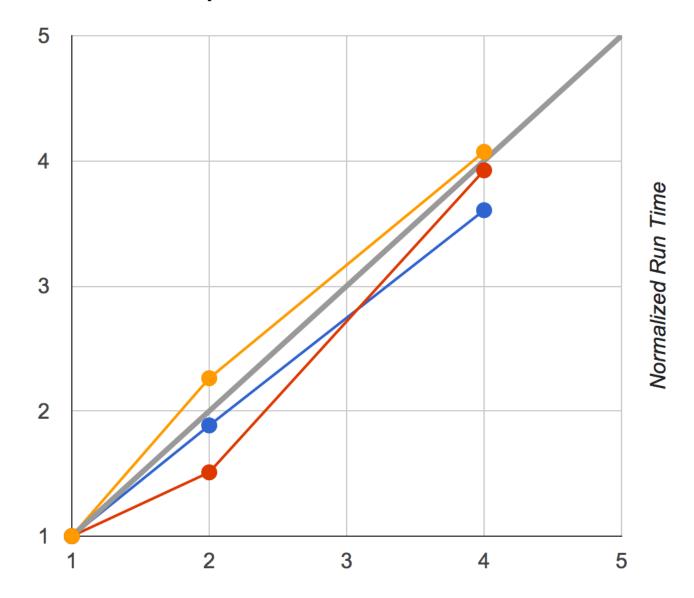
Algorithm	Run time (sec) for 50GB (800M rows)	Per-Core Throughput (MB/sec-core)	Per-Machine Throughput (MB/s)
bigr.lm (ridge)	3.04	130	1,040
bigr.lm	4.05	102	816
bigr.lm.gd	12.2	32	256
bigr.glm.gd	24.5	16	128
bigr.glm	36.1	11	88
bigr.kmeans	335	1.2	9.6

**pAnalytics** performance on building machine learning models with cluster Adatao16 (m3.2xlarge) on a 50GB data set of 5 features and 800 million rows. (Gradient descent algorithms are over 5 iterations)

Algorithm	Run time (sec) for 1.1 TB dataset (1.6B rows)	Per-Core Throughput (MB/sec-core)	Per-Machine Throughput (MB/s)
bigr.lm (ridge)	70.9	130	1,040
bigr.lm	74.9	123	984
bigr.lm.gd	127	72.8	582
bigr.glm.gd	145	63.6	509

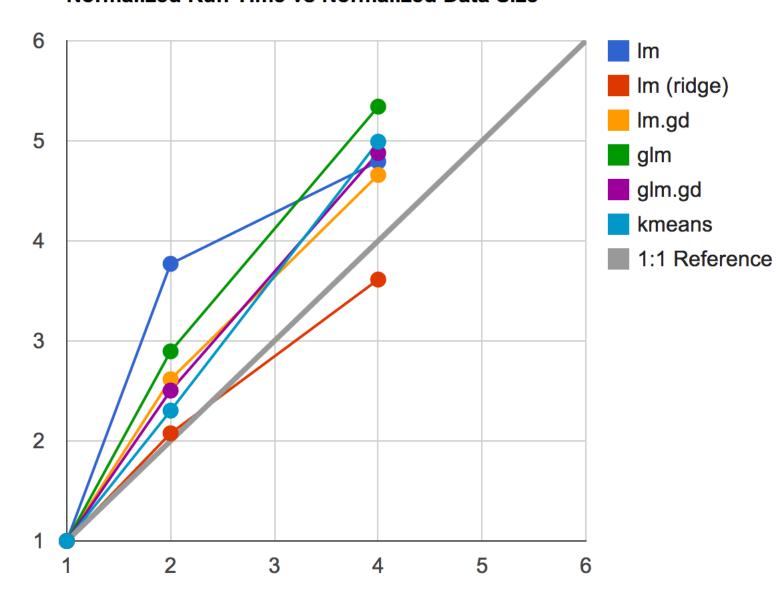
**pAnalytics** performance on building machine learning models with cluster Adatao40 (m3.2xlarge) on a 1.1 TB data set of 40 features and 1.6 billion rows. (Gradient descent algorithms are over 5 iterations)

#### **Normalized Speed vs Normalized Core Count**



Normalized Core Count (actual: 32 to 128)

#### Normalized Run Time vs Normalized Data Size



Normalized Data Size

# Data Intelligence for All



#### **Business Users**



PINSIGHI

#### Fast & Easy Business Analytics

Natural Language

Beautiful Web Ul

Data Scientists & Engineers



**PANALYFICS** 

Big & Fast Data Science

R, Python, REST API

Data Mining & ML



Thanks for contributing to the Spark Community!

#### Linear Regression, throughput vs. data size

