



**CLOUD NATIVE  
COMPUTING  
FOUNDATION**

# Machine learning and AI in the datacenter and how it will affect you

Nick Chase

Head of Content, Mirantis

Editor in Chief, Open Cloud Digest

Author, *Machine Learning for Mere Mortals*

# Who am I?

## Nick Chase



- Long-time programmer
- Head of Technical and Marketing Content for Mirantis
- Editor in Chief of *Open Cloud Digest*
- Author of *Machine Learning for Mere Mortals*
- Absolutely **NOT** a math major
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# Shhhhhh.....

Turns out, this is not as tough as you think.

# Special Bonus

- Free early access to *Machine Learning for Mere Mortals* for live attendees

[http://bit.ly/machine\\_learning\\_course](http://bit.ly/machine_learning_course)

# Agenda

- What is machine learning?
- Types of machine learning
- Problems facing the datacenter and how ML can solve them
- You, yes, YOU, can do this (and I'll prove it)
- Intelligent Delivery datacenter and how to start
- Q&A



# Introduction

Just what is machine learning,  
anyway?

# What is machine learning?

Machine learning is the study of enabling a computer to perform in situations it may not have seen before.

# Where can we see ML today?

- Analytics
- Advertising
- Bots
- Pattern recognition
- NLP
- Search



# Where can we see ML today?

- Analytics
- Advertising
- Bots
- Pattern recognition
- NLP
- Search
- Image recognition
- Video analysis
- Threat detection
- Medicine
- Self driving cars

# Lots of hype

- Existing companies

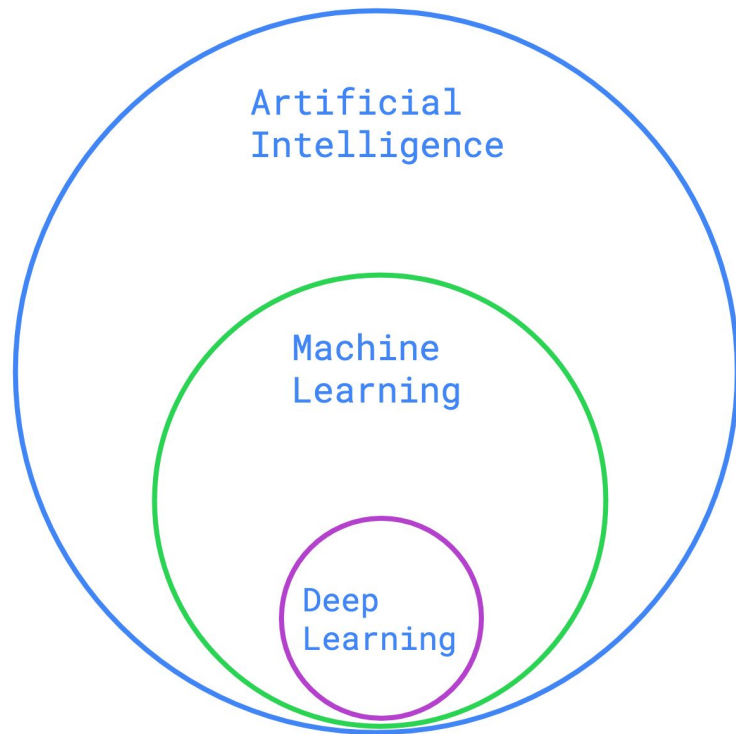
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- [Amazon.com](#) (1252)
- [JPMorgan Chase](#) (733)
- [Goldman Sachs](#) (674)
- [Microsoft](#) (580)
- [Apple](#) (322)
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- [IBM](#) (101)

# Lots of hype

- New companies (<http://bit.ly/2E2qFoU>)
- Newly accessible
  - Tensorflow
  - SciKit-Learn
  - etc.

# Machine learning vs artificial intelligence



# What does ML have to do with "cloud native"?

- Dynamic environments

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- Dynamic environments
- High-velocity management
- Continuous automation
- Observability
- Efficiency

# What can ML do for the datacenter?

Continuous Delivery infrastructure

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Continuous Delivery infrastructure



Intelligent Delivery infrastructure

# Intelligent Delivery infrastructure

- Defined architecture

# Intelligent Delivery infrastructure

- Defined architecture
- Flexible but controllable infrastructure

# Intelligent Delivery infrastructure

- Defined architecture
- Flexible but controllable infrastructure
- Intelligent oversight

# Intelligent Delivery infrastructure

- Defined architecture
- Flexible but controllable infrastructure
- Intelligent oversight
- Secure footing



# Types of Machine Learning



# Supervised learning

Training a computer to understand data it hasn't seen by teaching it about data that has already been labeled

# Supervised learning: Examples

- Image recognition

# Supervised learning: Examples

- Image recognition
- Speech recognition

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- Image recognition
- Speech recognition
- Medical outcome prediction

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- Image recognition
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- Spam filtering

# Supervised learning: Examples

- Image recognition
- Speech recognition
- Medical outcome prediction
- Spam filtering
- Stock market/crypto pricing

# Unsupervised learning

Letting the computer discover patterns and relationships in unlabeled data.

# Unsupervised learning: example

- Clustering



# Unsupervised learning: example

- Clustering
  - Customers into categories

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  - Chemical compounds

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- Anomaly detection

# Unsupervised learning: example

- Clustering
  - Customers into categories
  - Chemical compounds
- Anomaly detection
- Signal separation

# Reinforcement learning

A combination of supervised and unsupervised learning, where the computer is given an objective and attempts to reach that objective, discovering tactics as it goes along.

# Reinforcement learning: example

- Just about anything humans do

# Reinforcement learning: example

- Just about anything humans do
- Self-driving cars

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- Strategy development



# Reinforcement learning: example

- Just about anything humans do
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- Strategy development
  - Games

# Reinforcement learning: example

- Just about anything humans do
- Self-driving cars
- Strategy development
  - Games
  - Scheduling



# Problems Facing the Datacenter

# Configuration: potential issues

- Self-service

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- Self-service
- Microservices architectures

# Configuration: potential issues

- Self-service
- Microservices architectures
- Dynamic environments and drift issues

# Configuration: how ML/AI can help

- App configuration based on past history

# Configuration: how ML/AI can help

- App configuration based on past history
- Reinforcement learning to retroactively improve configuration



# Performance optimization: Potential issues

- Lots of factors in play

# Performance optimization: Potential issues

- Lots of factors
- Dynamic environments create a changing landscape

# Performance optimization: how ML/AI can help

- Analyze multiple factors

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- Analyze multiple factors
- Predict future load and move or scale workloads based on patterns

# Performance optimization: how ML/AI can help

- Analyze multiple factors
- Predict future load and move or scale workloads based on patterns
- Move workloads to closer geographical resources based on patterns

# Cost optimization: potential issues

- Complex environments

# Cost optimization: potential issues

- Complex environments
- Changing costs

# Cost optimization: potential issues

- Complex environments
- Changing costs
- Soaring storage requirements



# Cost optimization: how AI/ML can help

- Keep track of multiple environments and costs

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- Adapt to changing conditions

# Cost optimization: how AI/ML can help

- Keep track of multiple environments and costs
- Adapt to changing conditions
- Optimize what data is kept

# Fault detection: potential issues

- Hardware failure

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- Hardware failure
- Software configuration drift

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- Data corruption

# Fault detection: potential issues

- Hardware failure
- Software configuration drift
- Data corruption

No, I am **not** making that up: <http://bit.ly/cosmic-corruption>

# Fault detection: potential issues

- Hardware failure
- Software configuration drift
- Data corruption
- Resource overrun



# Fault detection: how AI/ML can help

- Autoscaling

# Fault detection: how AI/ML can help

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- Pattern recognition

# Fault detection: how AI/ML can help

- Autoscaling
- Pattern recognition
- Anomaly detection

# Security: potential issues

- DDOS

# Security: potential issues

- Distributed Denial Of Service attacks
- Advanced Persistent Threats

# Security: potential issues

- DDOS
- APT
- Inside jobs

# Security: how ML/AI can help

- Anomaly detection to find APTs and inside jobs

# Security: how ML/AI can help

- Anomaly detection to find APTs and inside jobs
- Pattern recognition to predict DDOS attacks before they're in full swing





# Try It Yourself

# Once upon a time...

# Once upon a time...

- Self-configuring

# Once upon a time...

- Self-configuring
- Self-optimizing

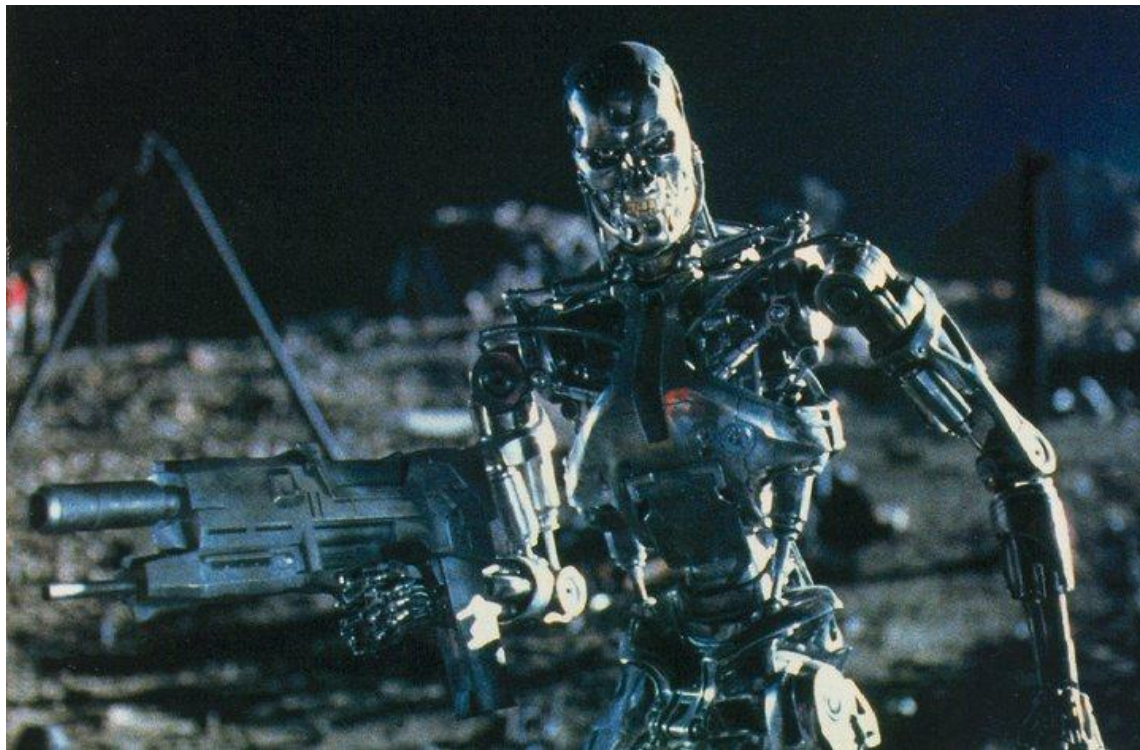
# Once upon a time...

- Self-configuring
- Self-optimizing
- Self-healing

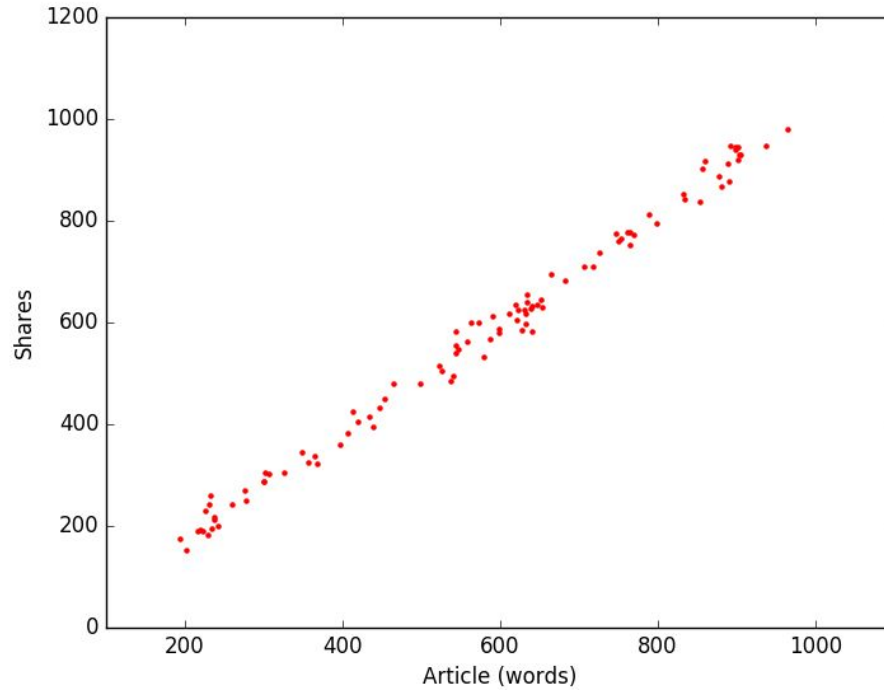
# Once upon a time...

- Self-configuring
- Self-optimizing
- Self-healing
- Self-protecting

Once upon a time...

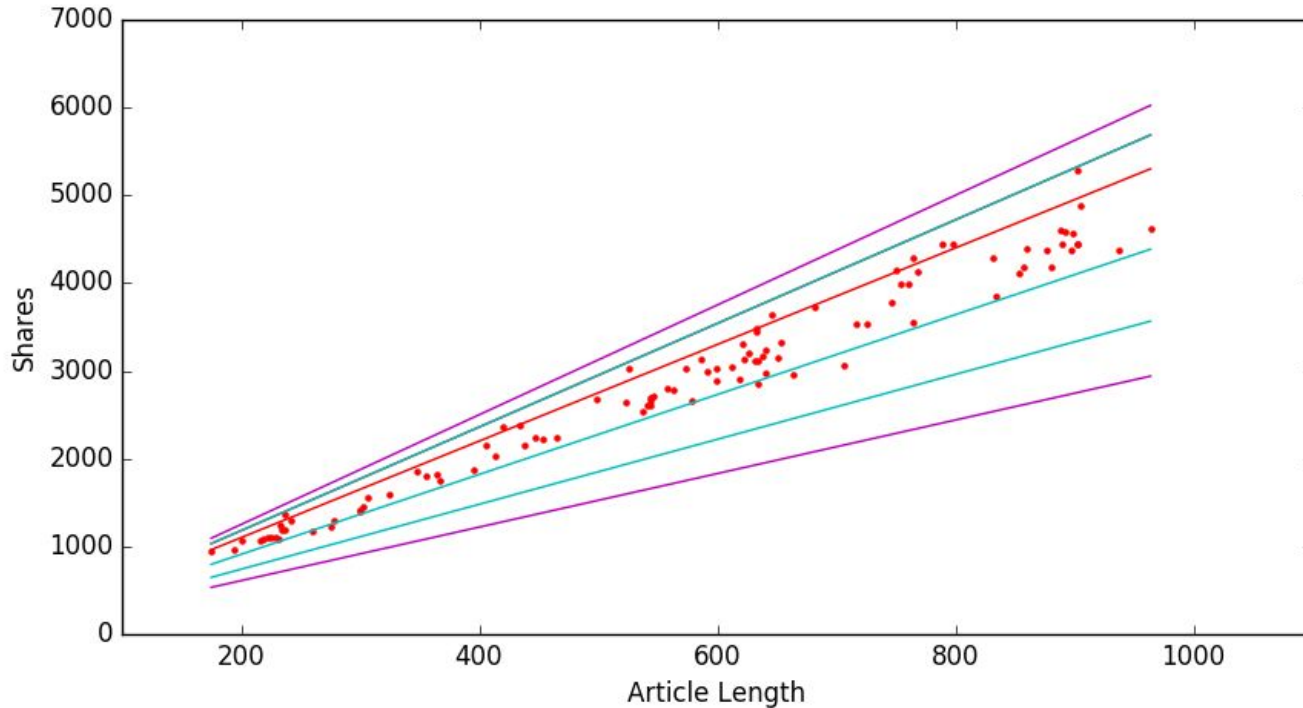


# What are we trying to do?

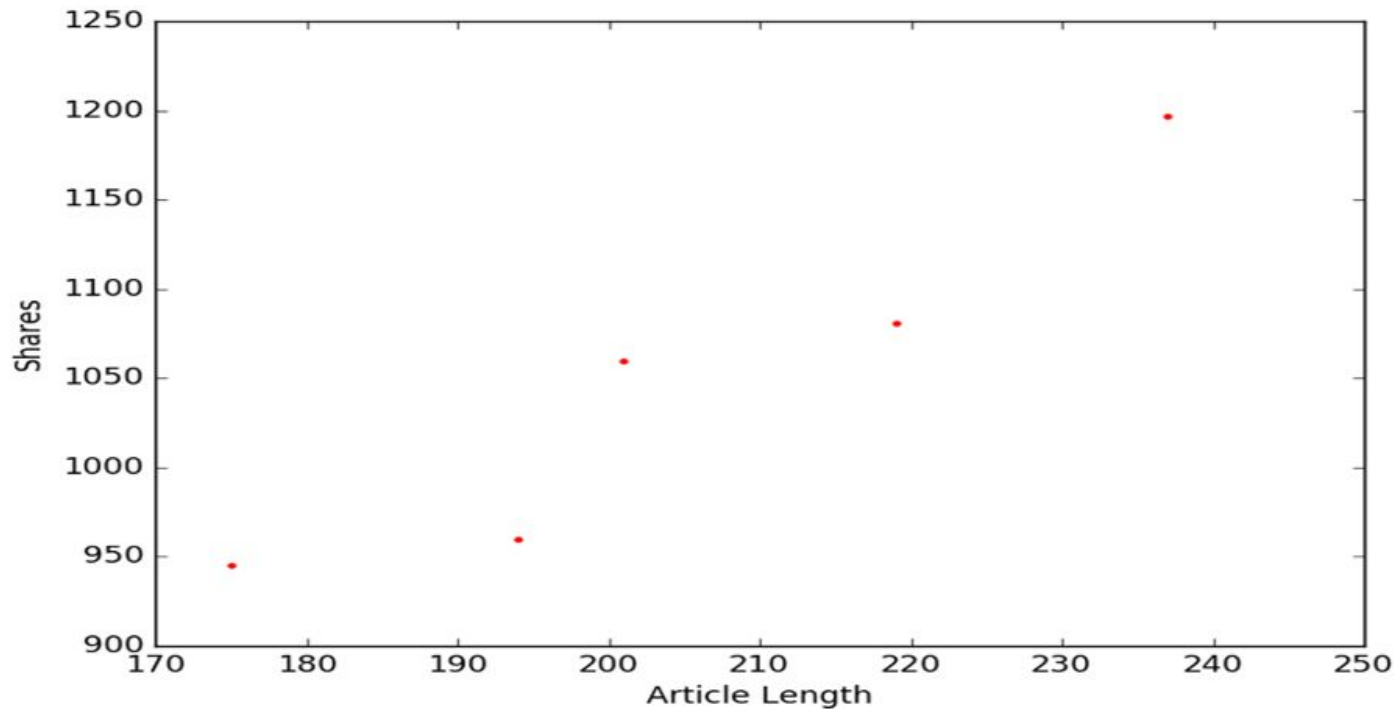




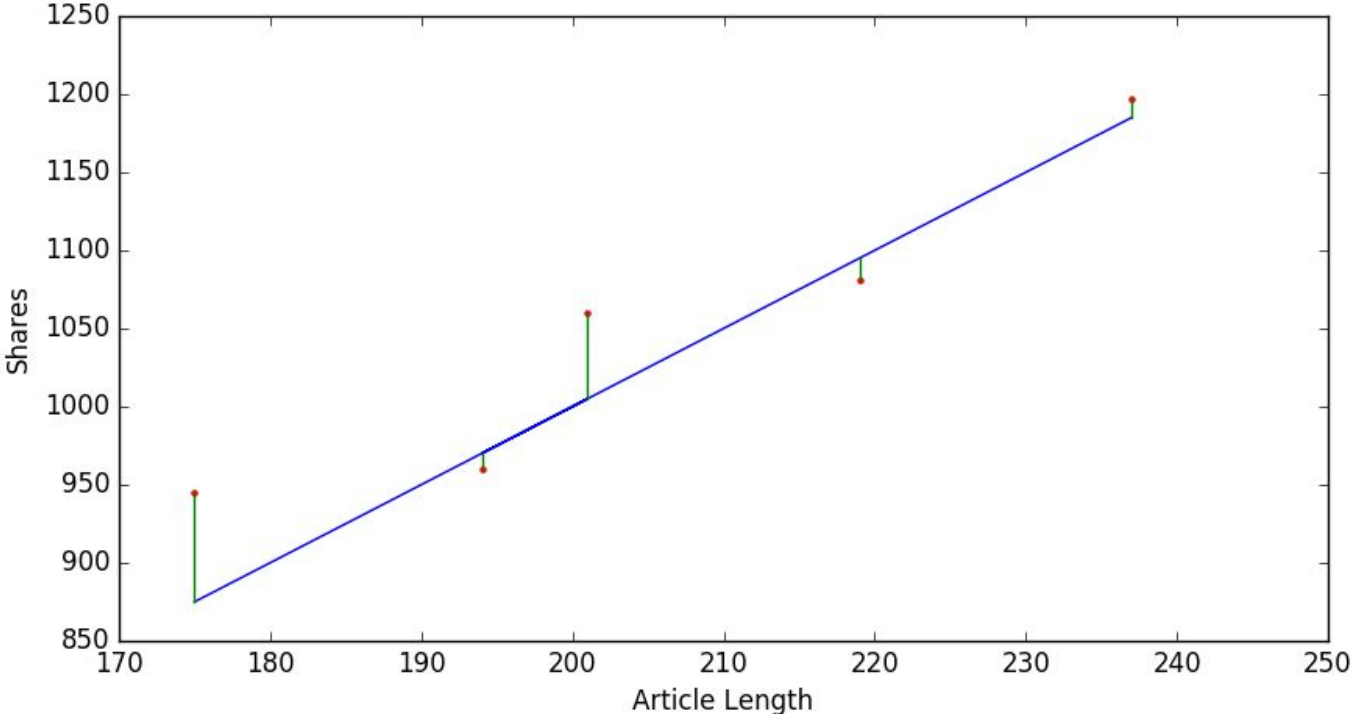
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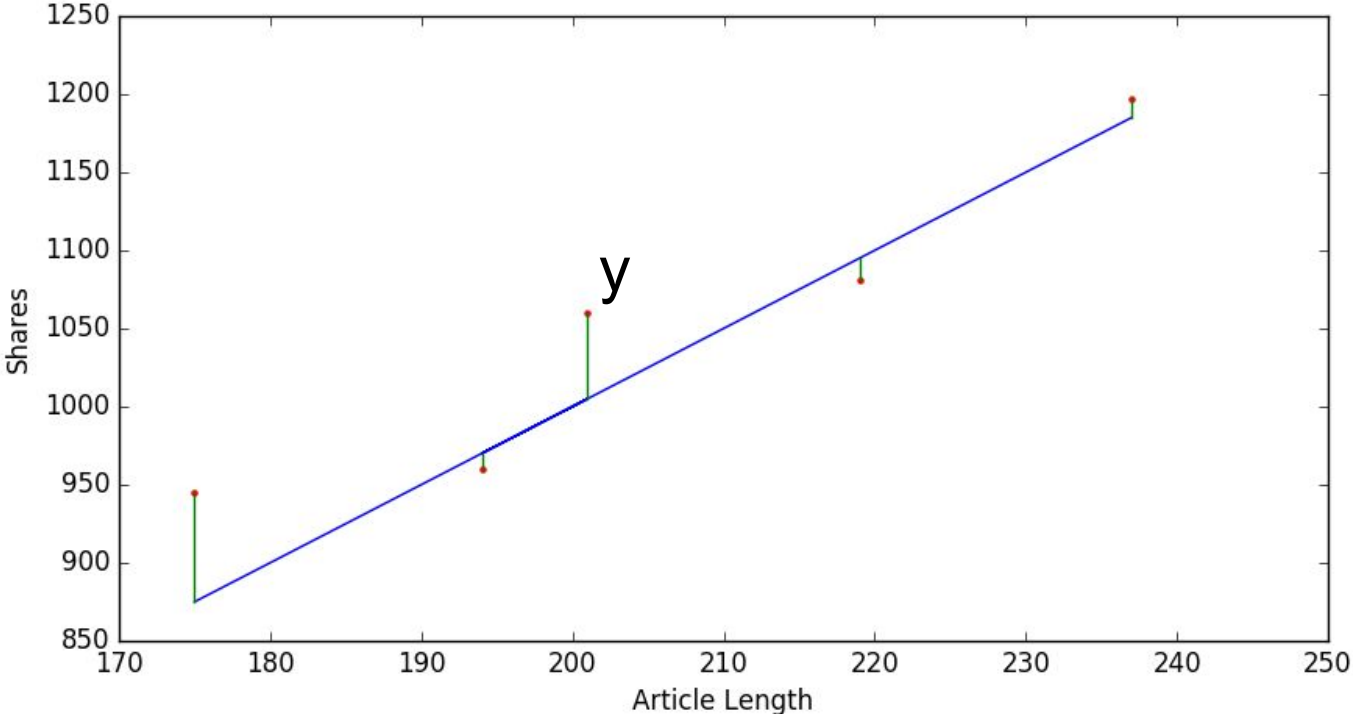
# Defining the error



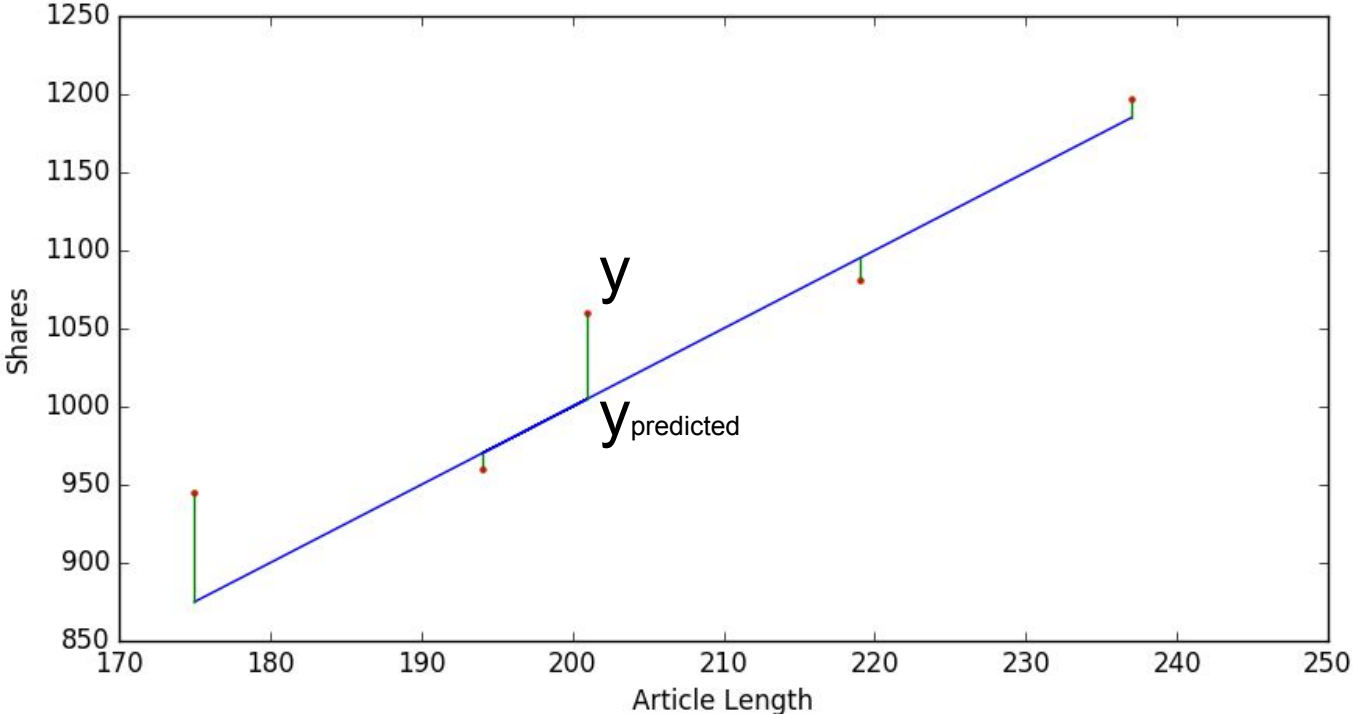
# Defining the error



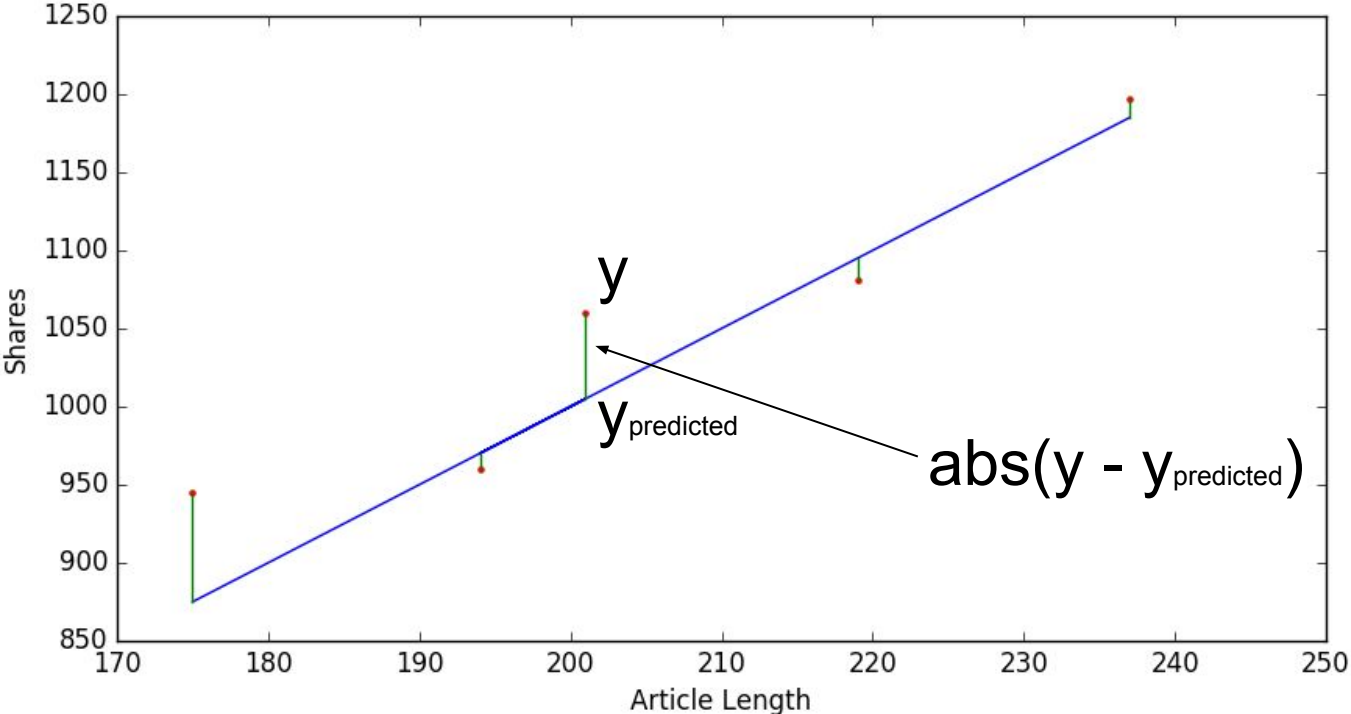
# Defining the error



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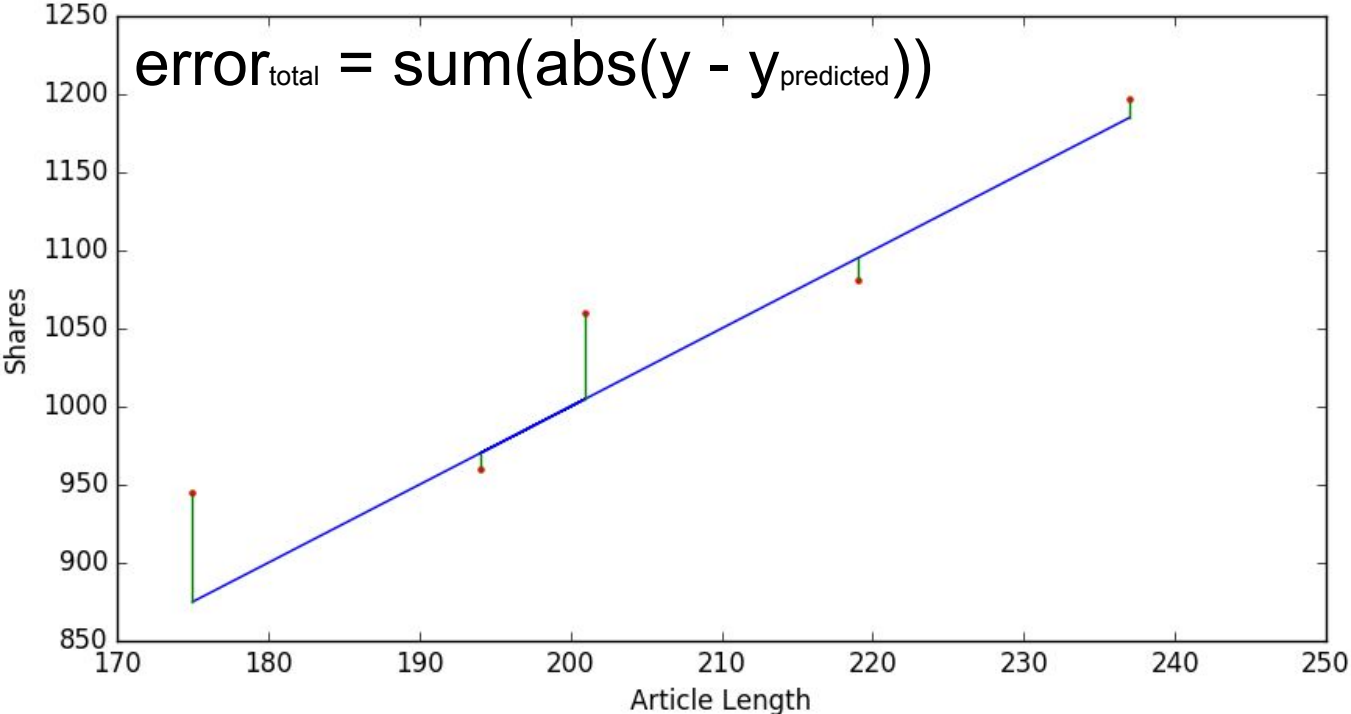


# Defining the error



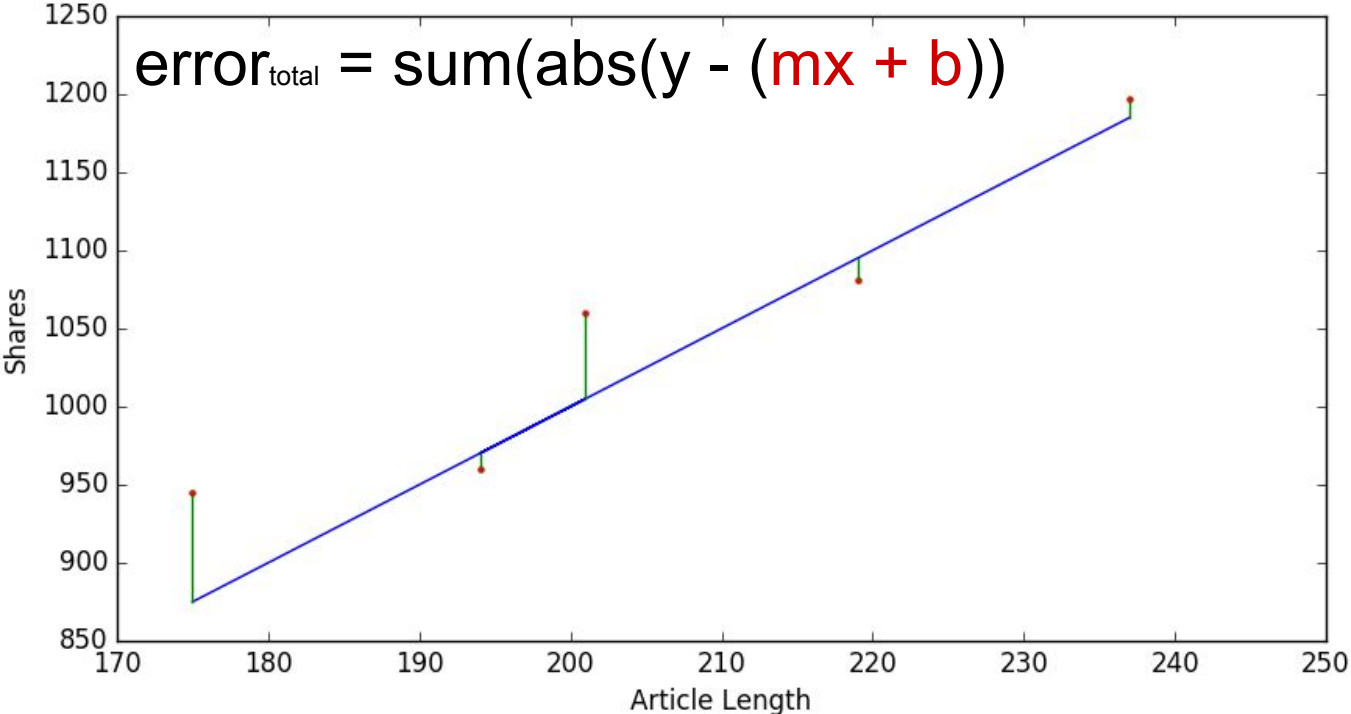
# Defining the error

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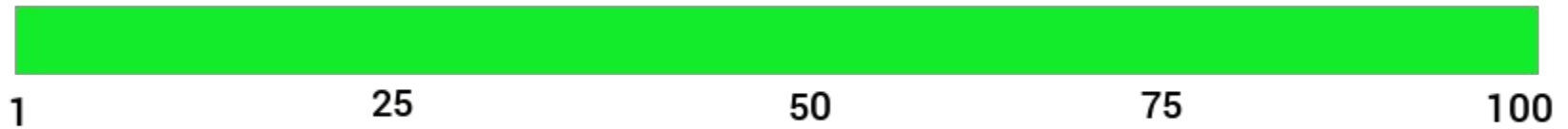
# Defining the error

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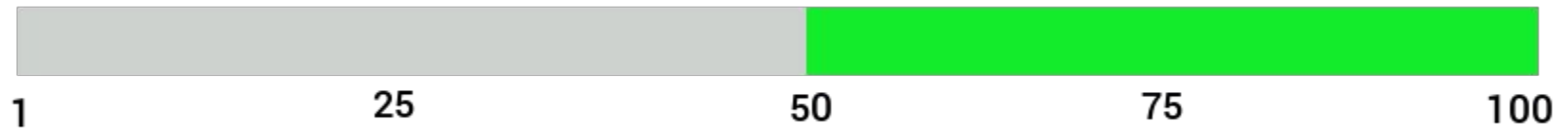




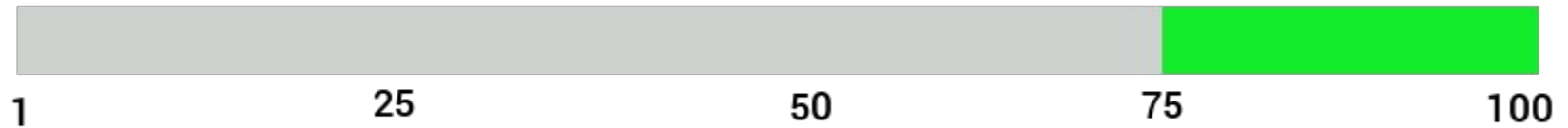
# Hot and cold



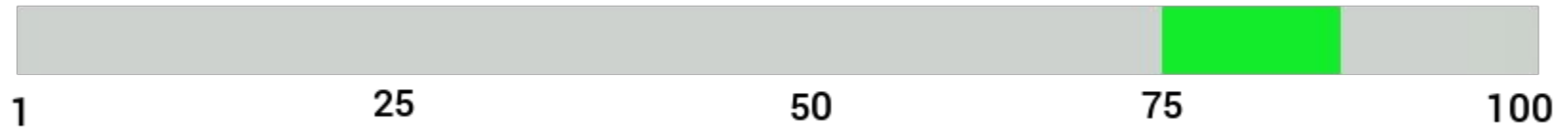
# Hot and cold



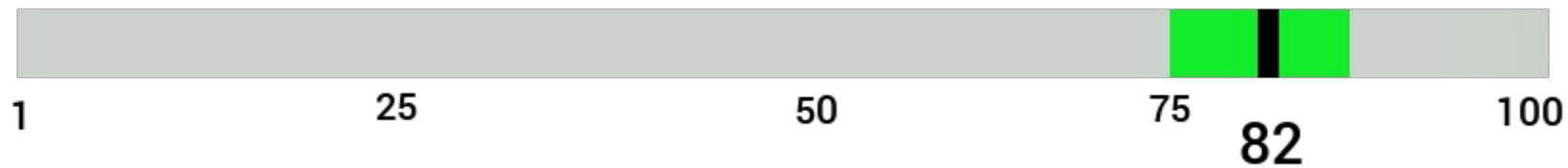
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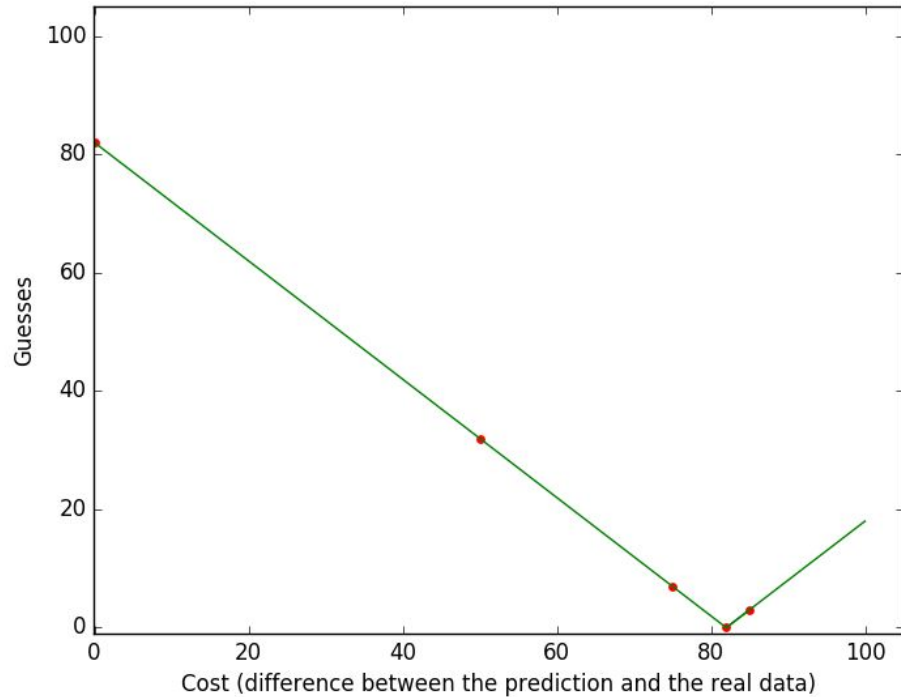
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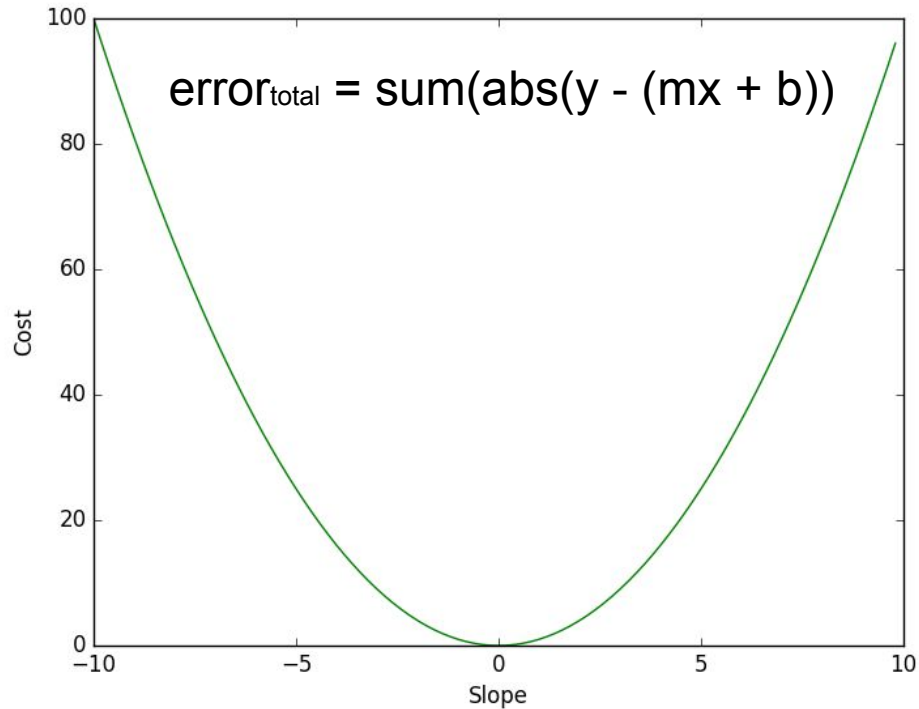
# Hot and cold



# Hot and cold



# Hot and cold



# Gradient descent

An iterative algorithm to find the minimum of the cost/error function and that parameters that achieve that minimum.



# Putting it together

- Data
- Python3
- Tensorflow
- Matplotlib (optional)
- Ubuntu 16.04

# What we'll do

- Read the data
- Create the model
- Define the error
- Define the optimizer
- Repeatedly run the optimizer until we get as close as we can

# DEMO

# The actual answer

$$y = .37x + 12$$

$$m = .37$$

$$b = 12$$

# DEMO



Where do we go from here?

# More practically

- Don't expect to go from B&W to VR in one step
  - Smaller steps in between

# More practically

- Don't expect to go from B&W to VR in one step
  - Smaller steps in between
- Start with analytics



# More practically

- Don't expect to go from B&W to VR in one step
  - Smaller steps in between
- Start with analytics
- Ultimately, know what you're trying to achieve



## Q&A



Thank You