

# **Microsoft Azure Machine Learning**

# INSTRUCTOR INFORMATION

Loy Vanich  
084 007 5544  
Line ID: laployv  
laploy@gmail.com  
www.laploy.com



# REPOSITORY AND NOTE SHARE

Repository

[github.com/laploy/ML](https://github.com/laploy/ML)

Note Share

[gist.github.com/laploy](https://gist.github.com/laploy)

(Azure ML Note for student)

# INTRODUCTION TO AZURE ML



What Bill have to say?

“A breakthrough in machine learning would be worth ten Microsofts,”

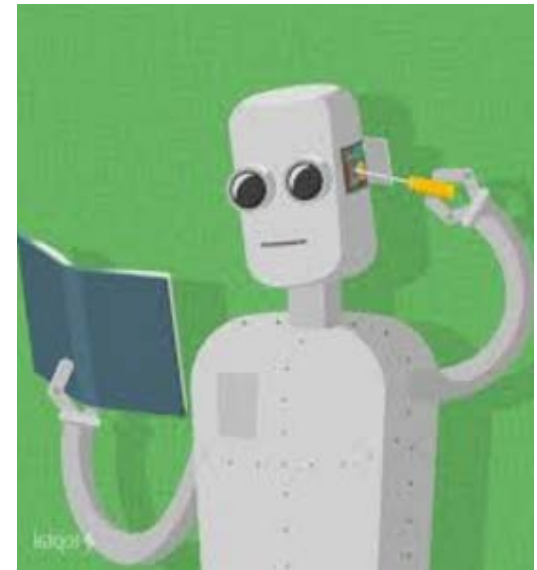
Bill Gates



What is Machine Learning?

## What is Machine Learning?

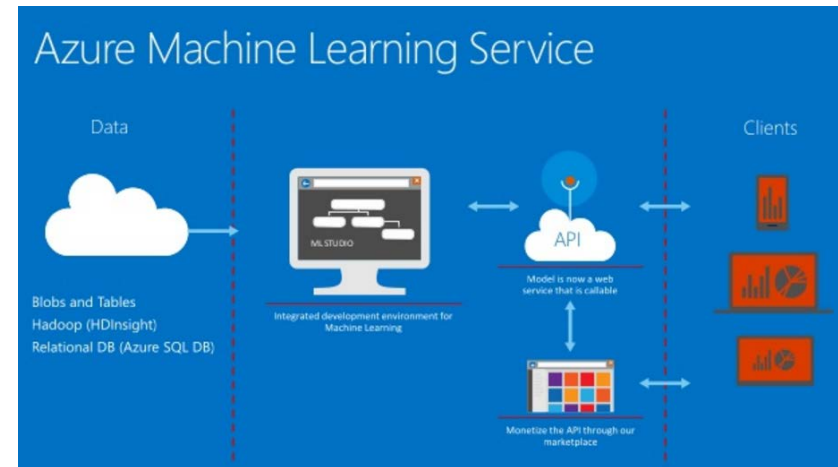
- Subfield of computer science
- Ability to learn without being explicitly programmed
- Algorithms that can learn from and make predictions on data
- Closely related to computational statistics
- Focuses on prediction-making
- Sometimes mixed with Data Mining (DM)
- Used complex models and algorithms
- Predictive analytics



## What is Azure ML?

### What is Azure ML?

- Pronounce = Air-Cher
- Part of Azure Cloud platform
- Part of Cortana Intelligence Suite
- Platform as a service (PaaS)
- Create systems that improve with experience
- Help turning data into software
- Help training with huge volumes of data
- Used to predict certain patterns, trends, and outcomes
- Predictive analytics is the underlying technology
- Use the past to predict the future



What is Azure ML Studio?

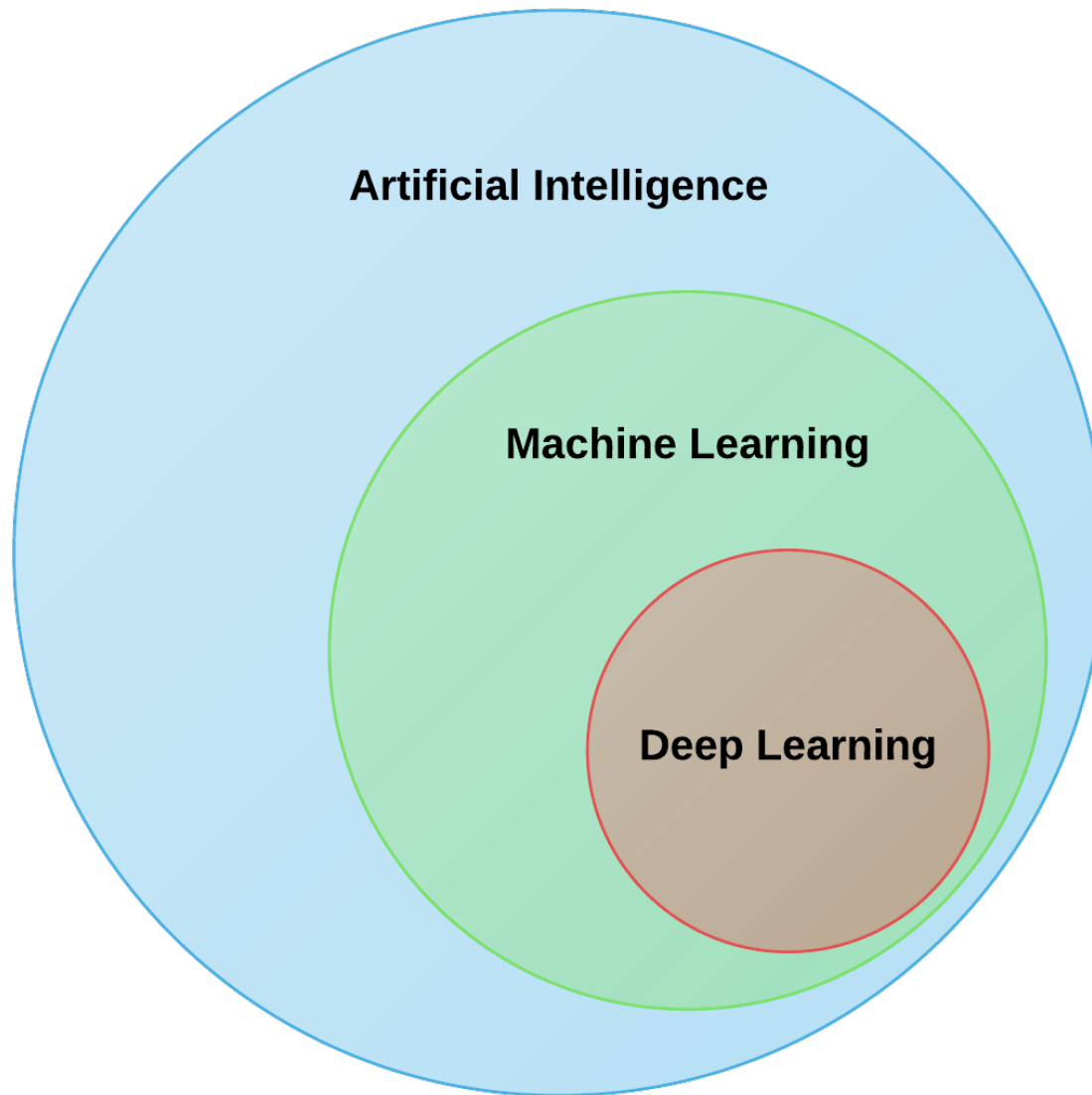
# Azure Machine Learning Studio

## What is Azure ML Studio?

- Is a collaborative tool
- Drag-and-drop
- Use to build, test, and deploy predictive analytics solutions
- Publishes models as web services
- Easily be consumed by custom apps or BI tools
- Is where data science, predictive analytics, cloud resources, and your data meet.

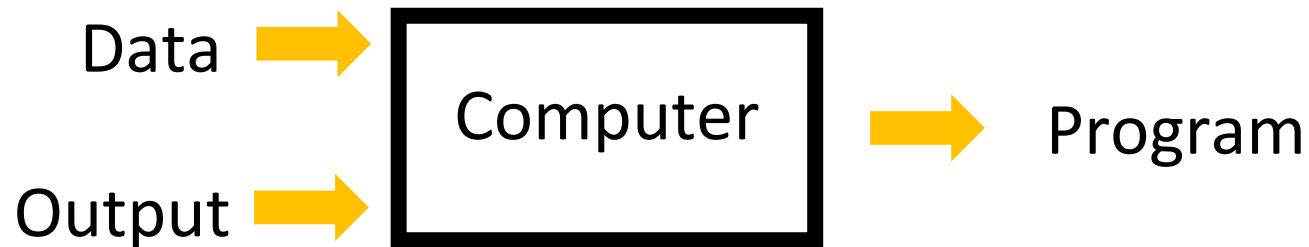
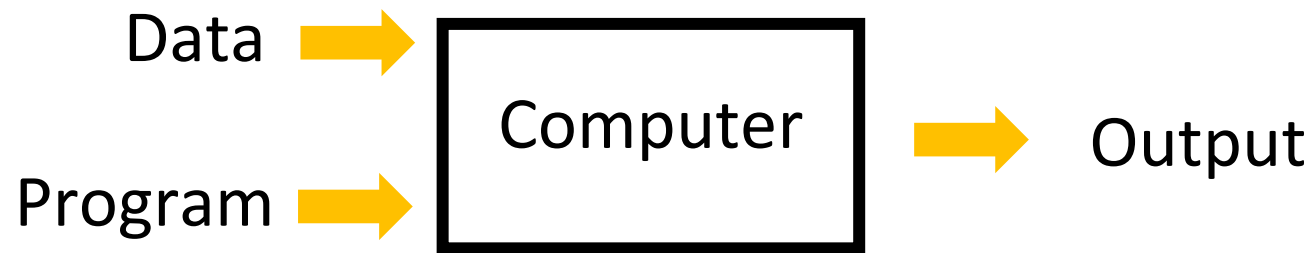


## Artificial Intelligence VS AI



## ML paradigm

### Traditional Programming



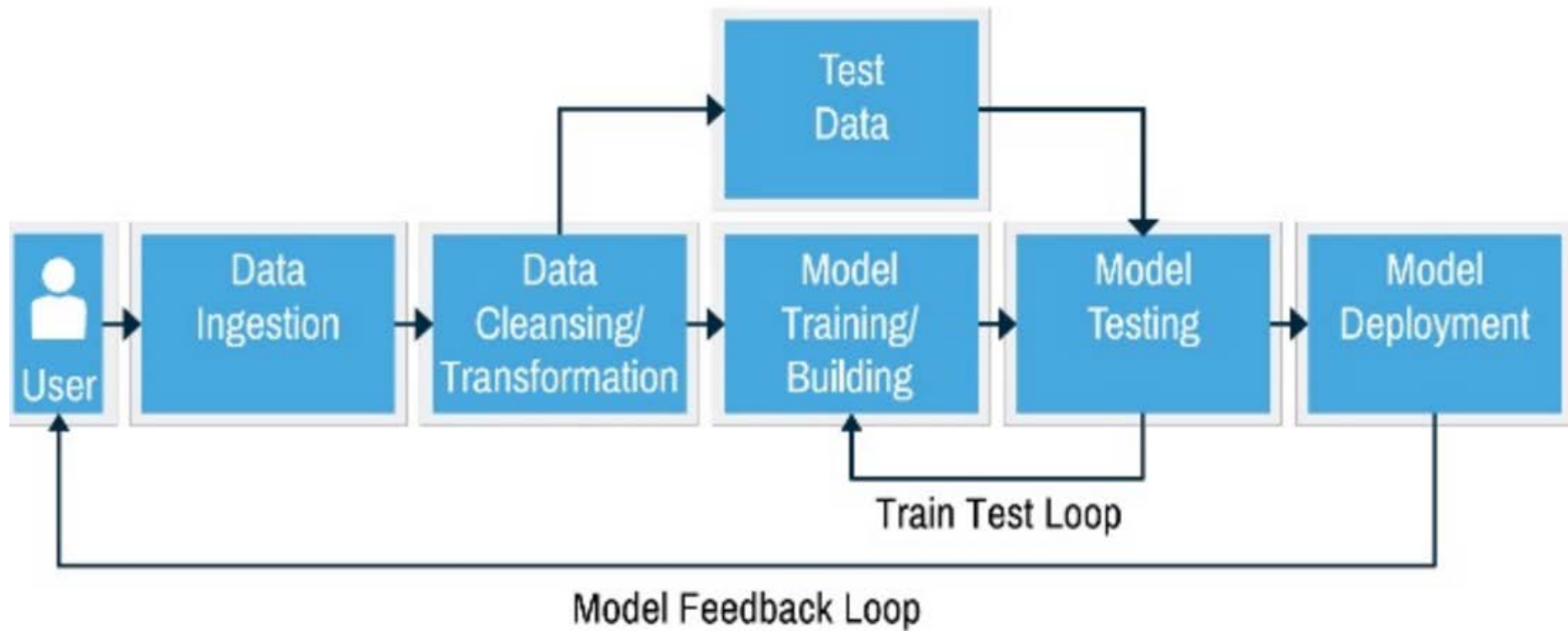
## Everyday examples of predictive analytics

### Everyday examples of predictive analytics

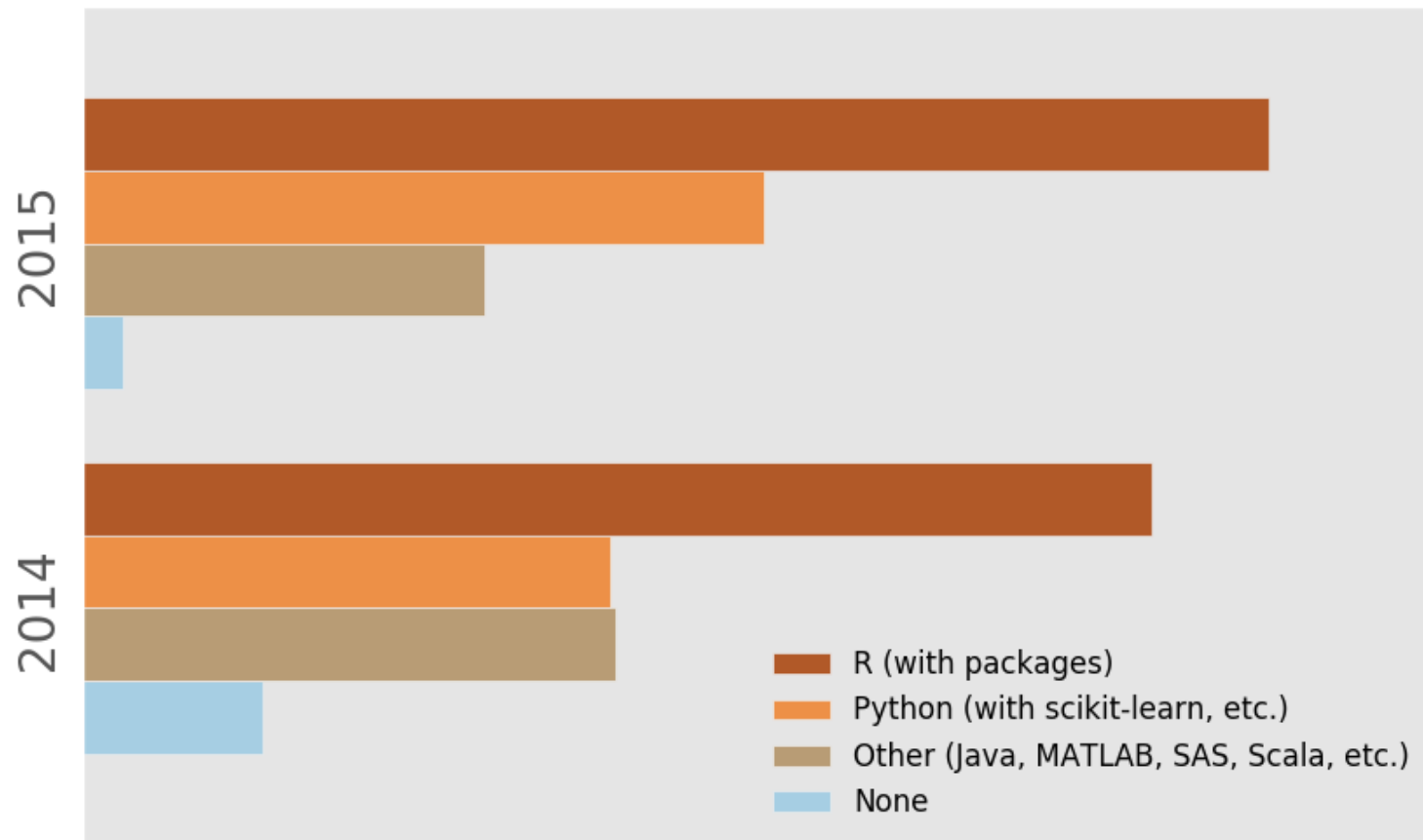
- Spam/junk email filters
- Mortgage applications
- Pattern recognition
- Life insurance
- Medical insurance
- Liability/property insurance
- Credit card fraud detection
- Airline flights
- Web search
- Predictive maintenance
- Health care



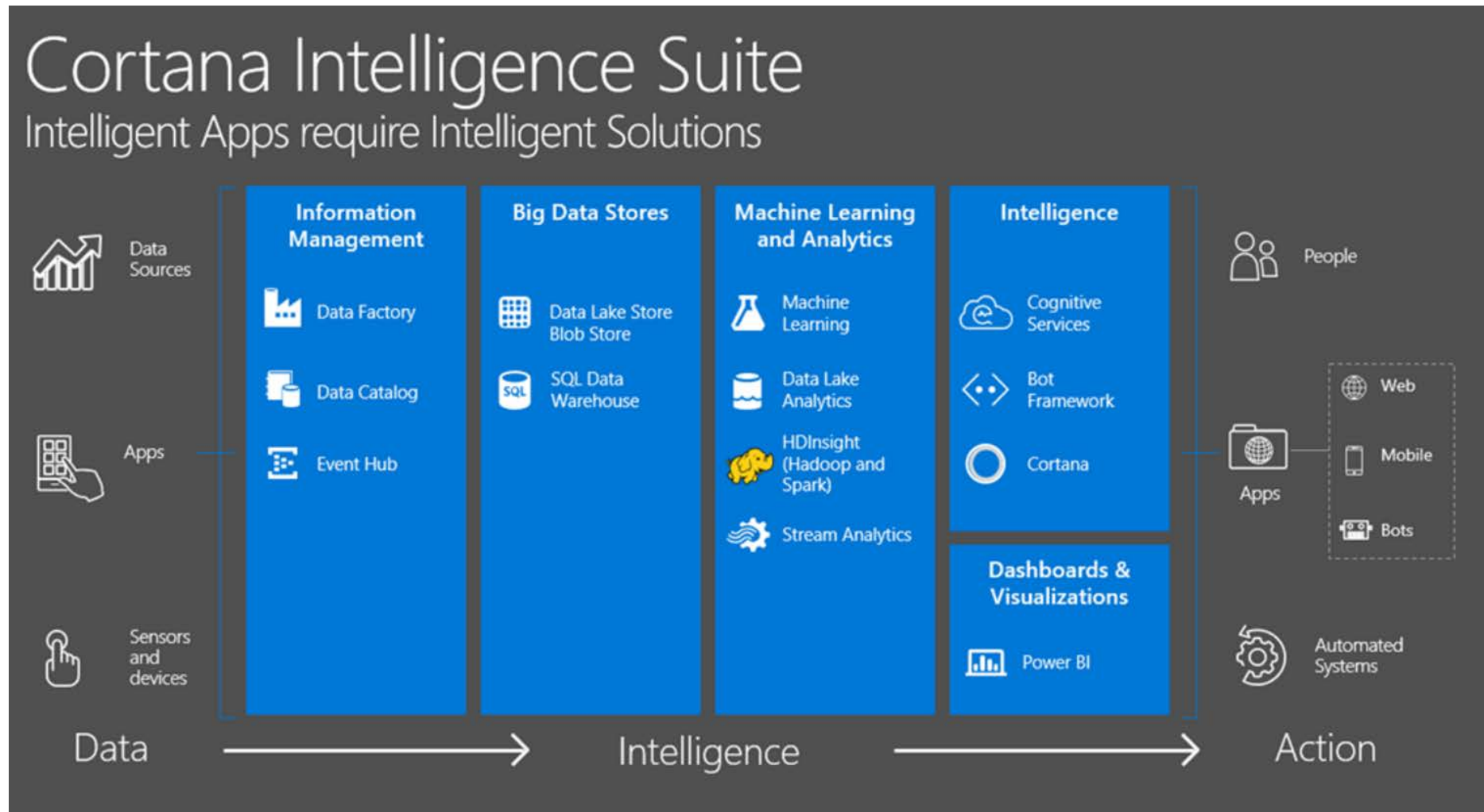
## Machine Learning work flow



## Machine learning language



## Cortana Intelligence Suite (CIS)



Use your own data to train models



TensorFlow



Cloud Machine  
Learning Engine

Ready to use Machine Learning models



Cloud  
Vision API



Cloud  
Speech API



Cloud  
Jobs API



Cloud  
Translation  
API



Cloud Natural  
Language API



Cloud Video  
Intelligence API



Coming  
soon

Amazon Machine Learning

# Introducing Amazon ML



Easy to use, managed machine learning service built for developers

Robust, powerful machine learning technology based on Amazon's internal systems

Create models using your data already stored in the AWS cloud

Deploy models to production in seconds

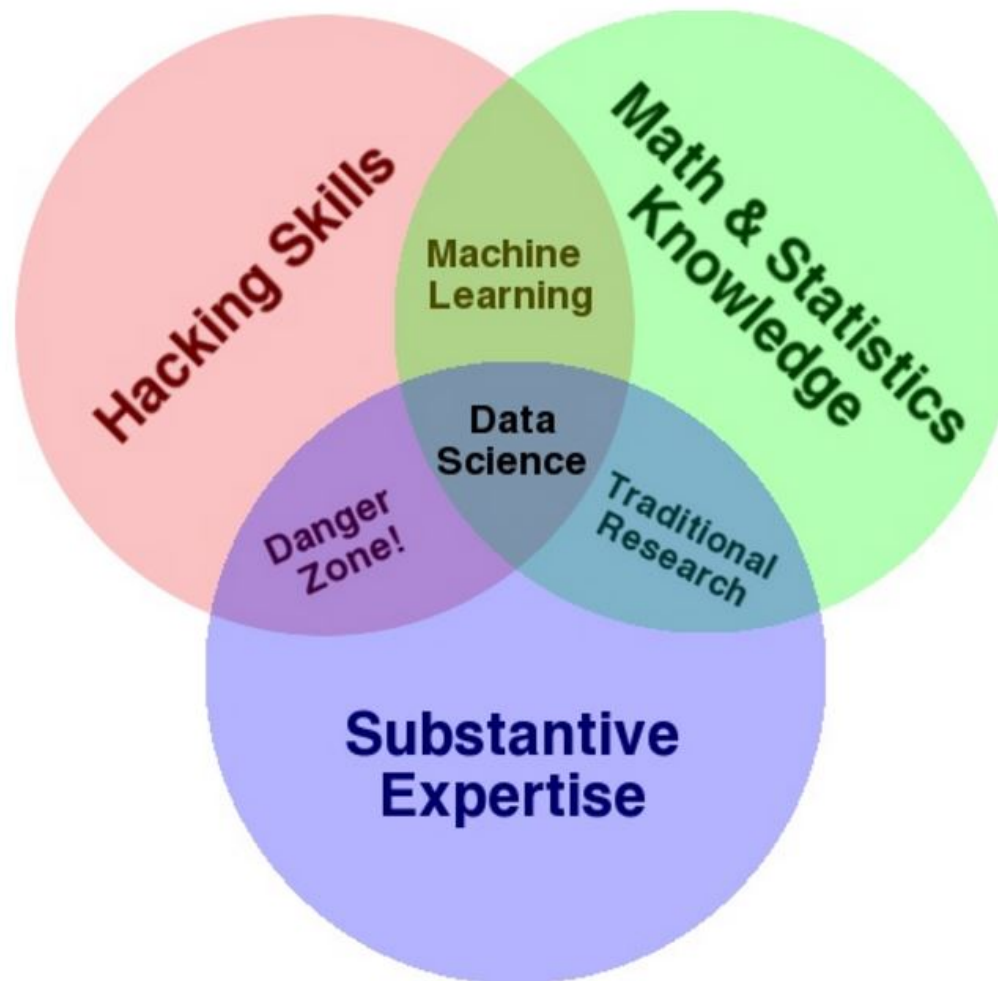




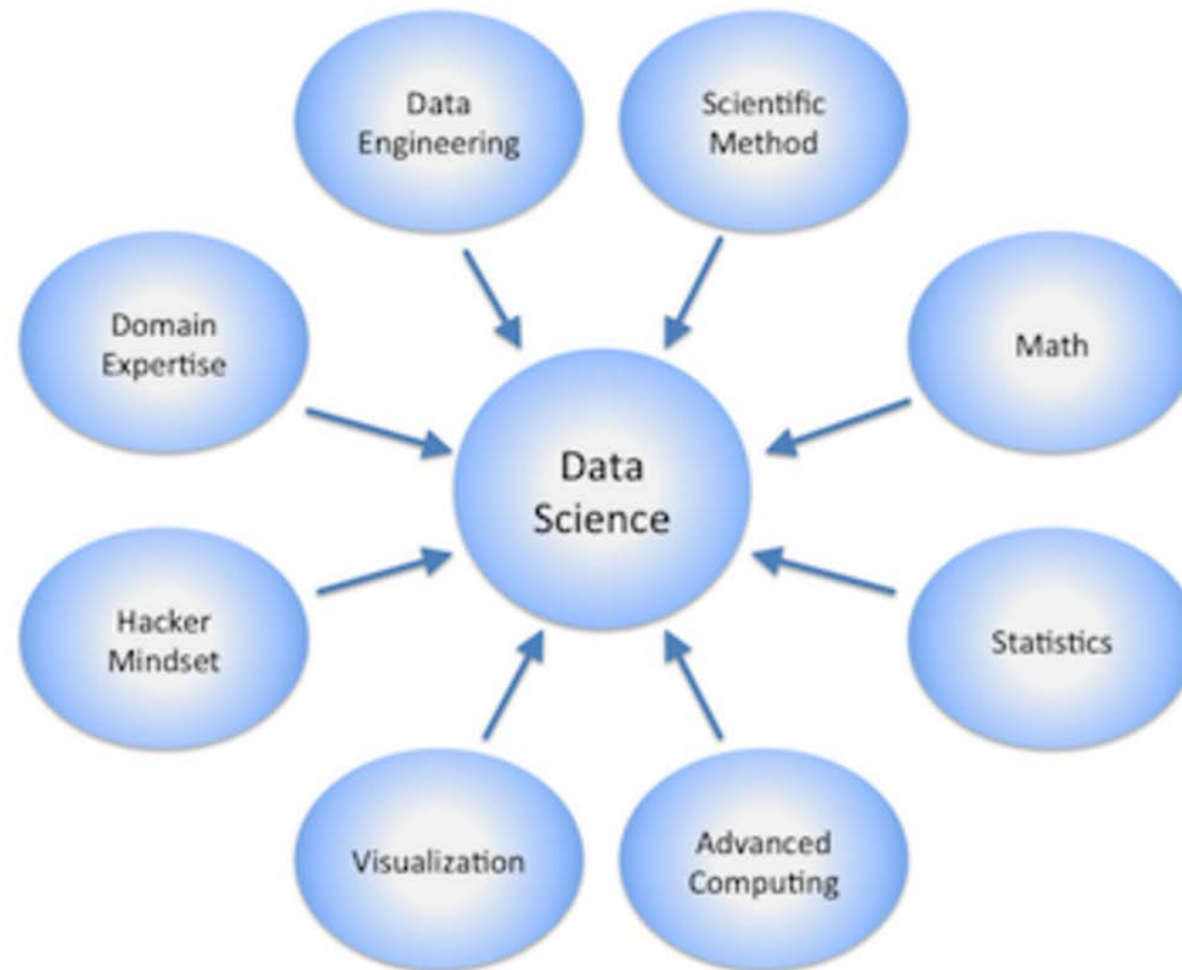
# DATA SCIENCE INTRODUCTION



## What is Data Science



## What is data science?



## Data scientist

### MATH & STATISTICS

- ☆ Machine learning
- ☆ Statistical modeling
- ☆ Experiment design
- ☆ Bayesian inference
- ☆ Supervised learning: decision trees, random forests, logistic regression
- ☆ Unsupervised learning: clustering, dimensionality reduction
- ☆ Optimization: gradient descent and variants

### DOMAIN KNOWLEDGE & SOFT SKILLS

- ☆ Passionate about the business
- ☆ Curious about data
- ☆ Influence without authority
- ☆ Hacker mindset
- ☆ Problem solver
- ☆ Strategic, proactive, creative, innovative and collaborative

### PROGRAMMING & DATABASE

- ☆ Computer science fundamentals
- ☆ Scripting language e.g. Python
- ☆ Statistical computing packages, e.g., R
- ☆ Databases: SQL and NoSQL
- ☆ Relational algebra
- ☆ Parallel databases and parallel query processing
- ☆ MapReduce concepts
- ☆ Hadoop and Hive/Pig
- ☆ Custom reducers
- ☆ Experience with xaaS like AWS

### COMMUNICATION & VISUALIZATION

- ☆ Able to engage with senior management
- ☆ Story telling skills
- ☆ Translate data-driven insights into decisions and actions
- ☆ Visual art design
- ☆ R packages like ggplot or lattice
- ☆ Knowledge of any of visualization tools e.g. Flare, D3.js, Tableau



## Glassdoor best job in 2016 - 2017

2016

**Data Scientist** (#1), **Tax Manager** (#2) and **Solutions Architect** (#3) stand out as the three Best Jobs in America for 2016. But which other jobs made the cut?

<https://www.glassdoor.com/blog/25-jobs-america-2016/>

2017

### 1 **Data Scientist**



**4.8 / 5**  
Job Score

**4.4 / 5**  
Job Satisfaction

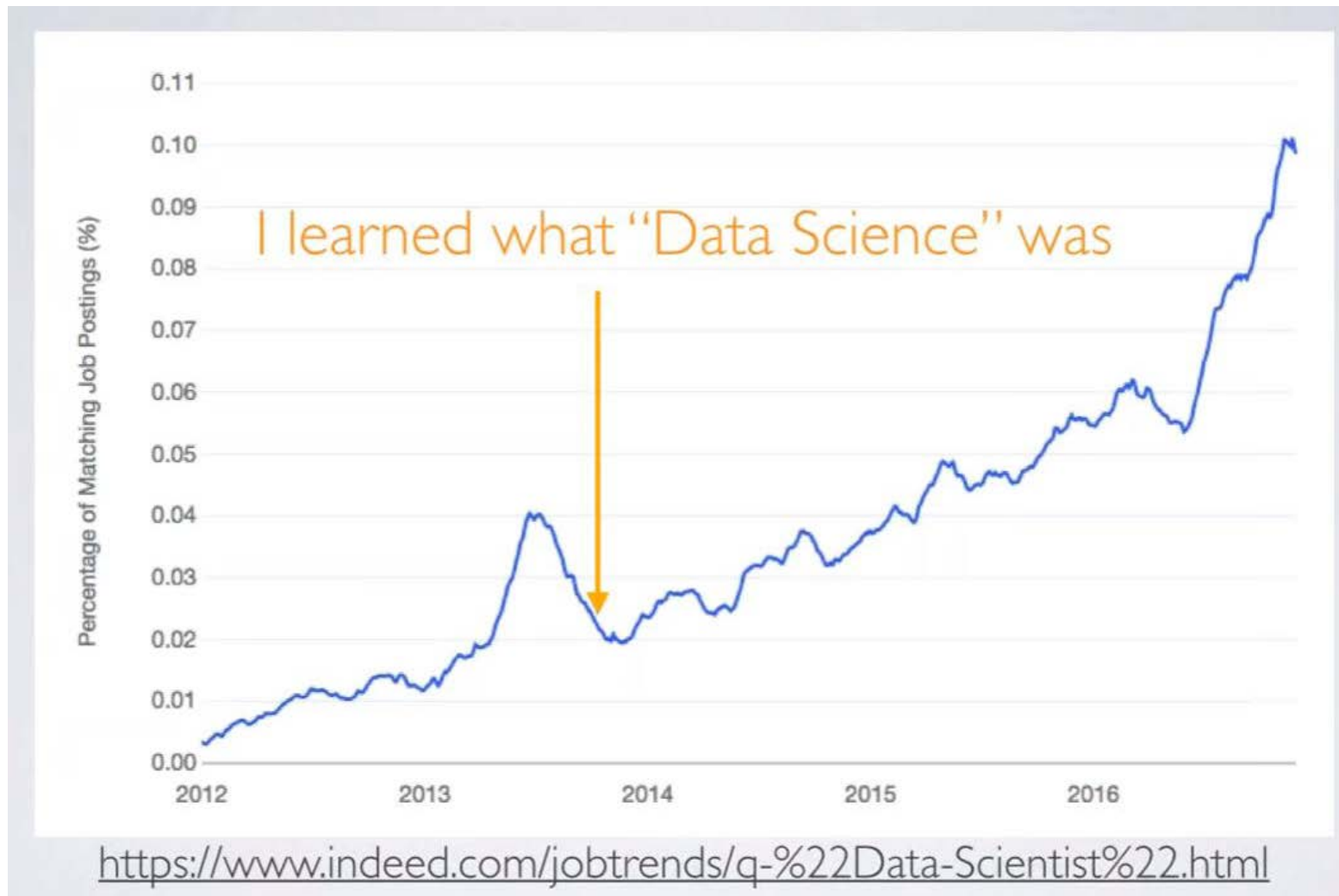
**\$110,000**  
Median Base Salary

**4,184**  
Job Openings

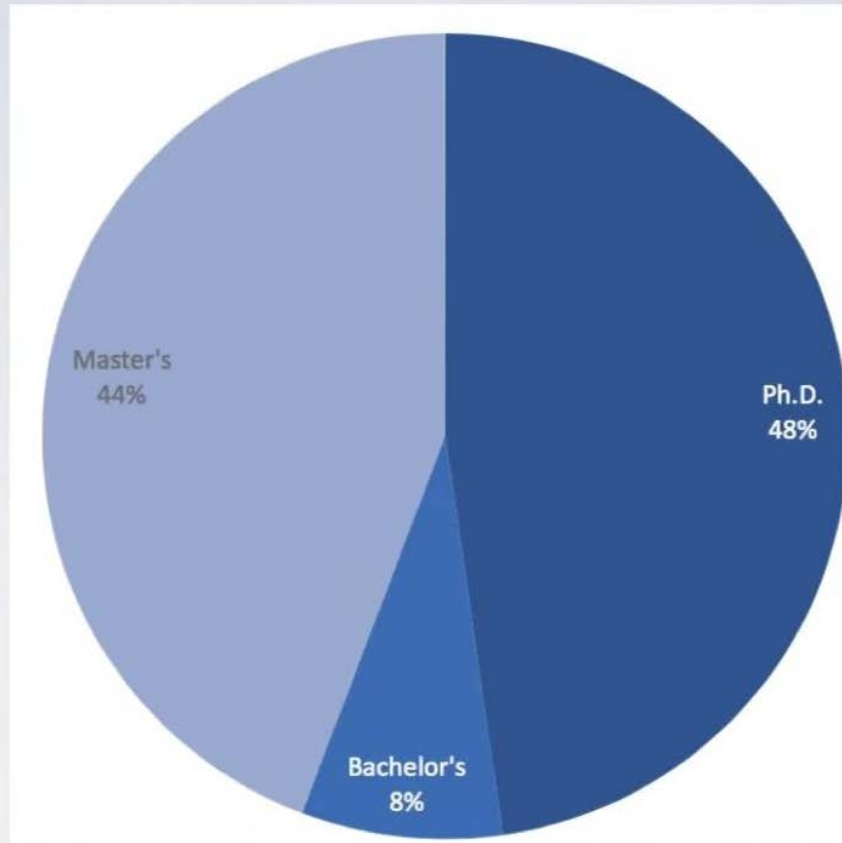
[View Jobs](#)

[https://www.glassdoor.com/List/Best-Jobs-in-America-LST\\_KQ0,20.htm](https://www.glassdoor.com/List/Best-Jobs-in-America-LST_KQ0,20.htm)

## Data science job trend



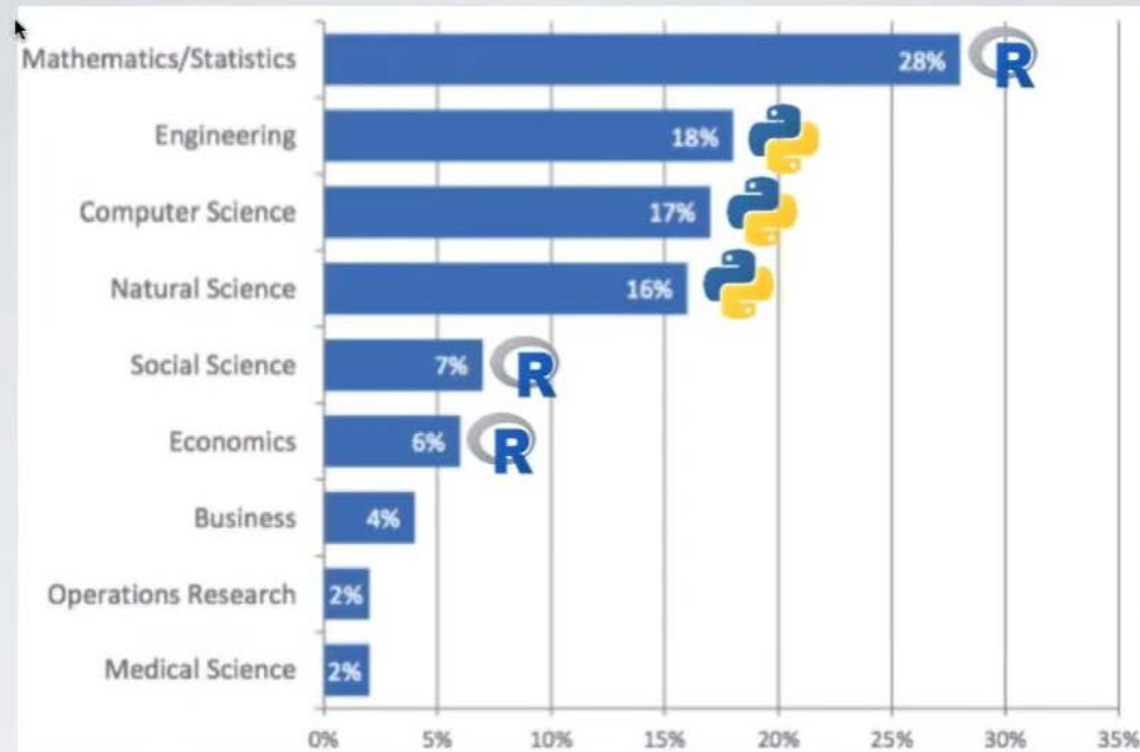
## Data Scientist education levels



Burtch Works 2016 Study, Data Scientist Education Levels

[http://www.burtchworks.com/files/2016/04/Burtch-Works-Study\\_DS-2016-final.pdf](http://www.burtchworks.com/files/2016/04/Burtch-Works-Study_DS-2016-final.pdf)

## Data science backgrounds



Burtch Works 2016 Study, Data Scientist Backgrounds

[http://www.burtchworks.com/files/2016/04/Burtch-Works-Study\\_DS-2016-final.pdf](http://www.burtchworks.com/files/2016/04/Burtch-Works-Study_DS-2016-final.pdf)



## Key topic to learn

I. Pick an **open-source** language well-designed for Data Science



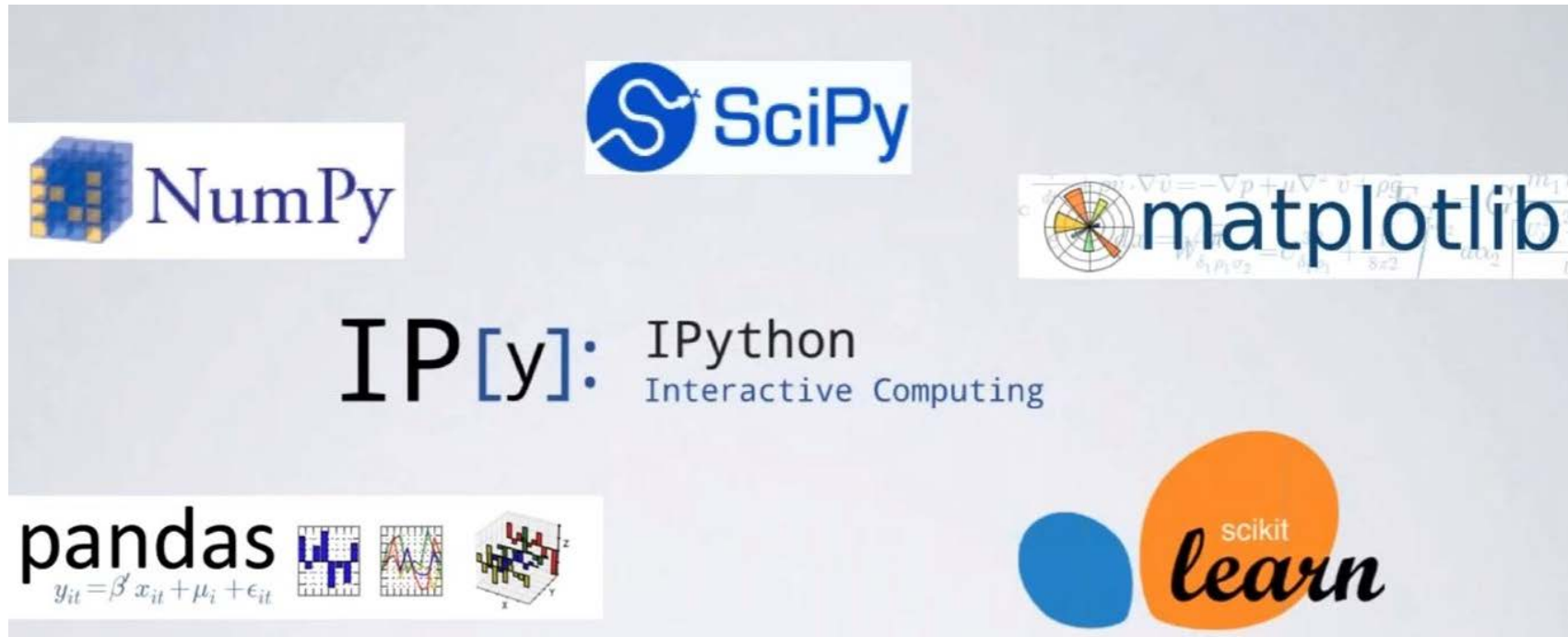
or



Python (my recommendation)

R (**if** you already know it well)

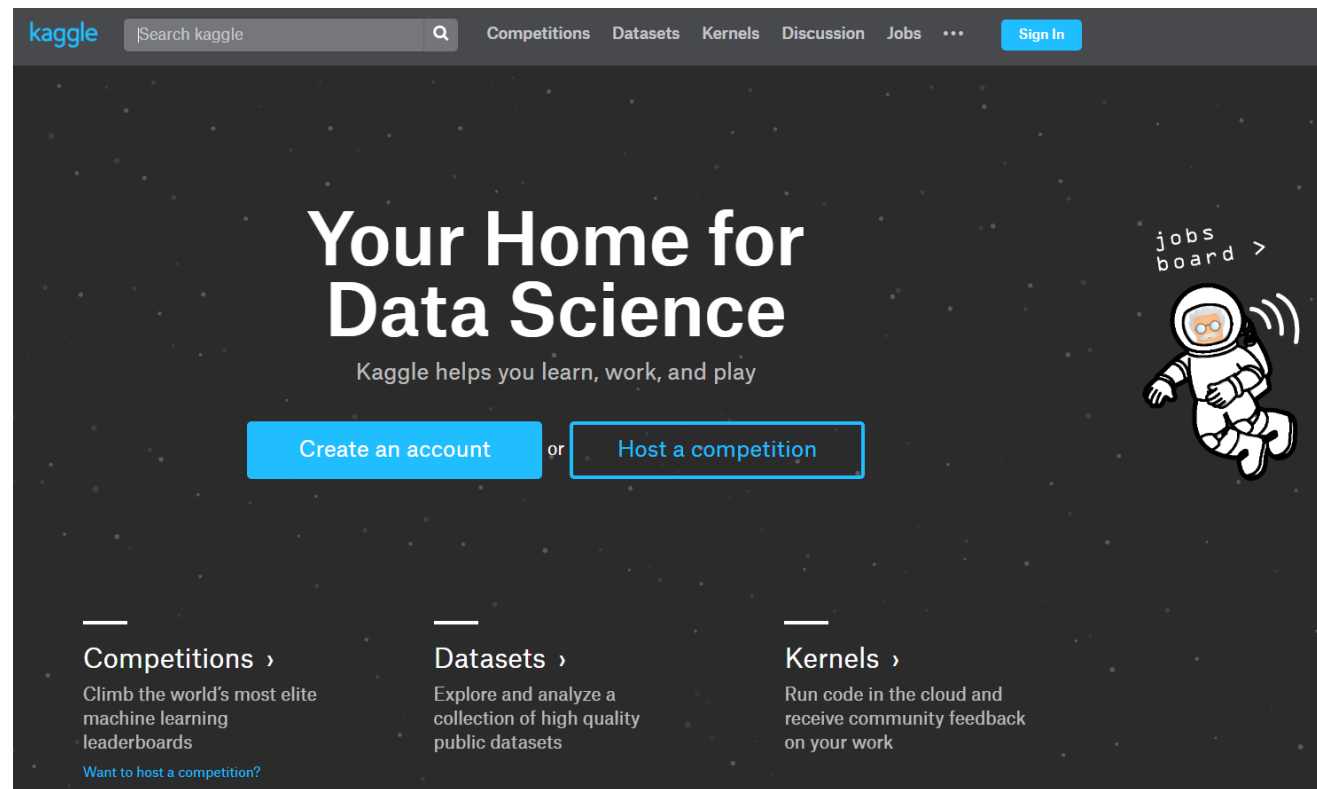
## Learn Python library stack



Go kaggle

- There are countless strategies that can be applied to any predictive modelling
- It is impossible to know at the outset which technique or analyst will be most effective
- Compete to produce the best models

kaggle™



# Launch Your Career in Data Science

A nine-course introduction to data science, developed and taught by leading professors.

## About This Specialization

Ask the right questions, manipulate data sets, and create visualizations to communicate results.

This Specialization covers the concepts and tools you'll need throughout the entire data science pipeline, from asking the right kinds of questions to making inferences and publishing results. In the final Capstone Project, you'll apply the skills learned by building a data product using real-world data. At completion, students will have a portfolio demonstrating their mastery of the material.

Created by:



DS compared to ML engineer

## How is a Machine Learning Engineer different from a Data Scientist?

### Data Scientist

- Trained to be strong in Data
- R, Python, MATLAB
- Data treatment
- Evaluate ML algorithm
- Evaluate ML module

### ML Engineer

- Trained to be strong in Coding
- C++, Java, C#
- Coding
- Change algorithm to code
- Create ML module

# DATA SCIENCE BASIC



## **In this session**

- The 5 questions data science answers
- Is your data ready for data science
- Ask a question you can answer with data
- Predict an answer with a simple model

The 5 questions data science answers

## The 5 questions data science answers

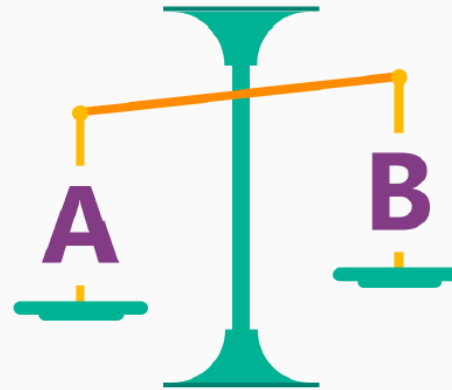
- Is this A or B?
- Is this weird?
- How much – or – How many?
- How is this organized?
- What should I do next?

The 5 questions data science answers



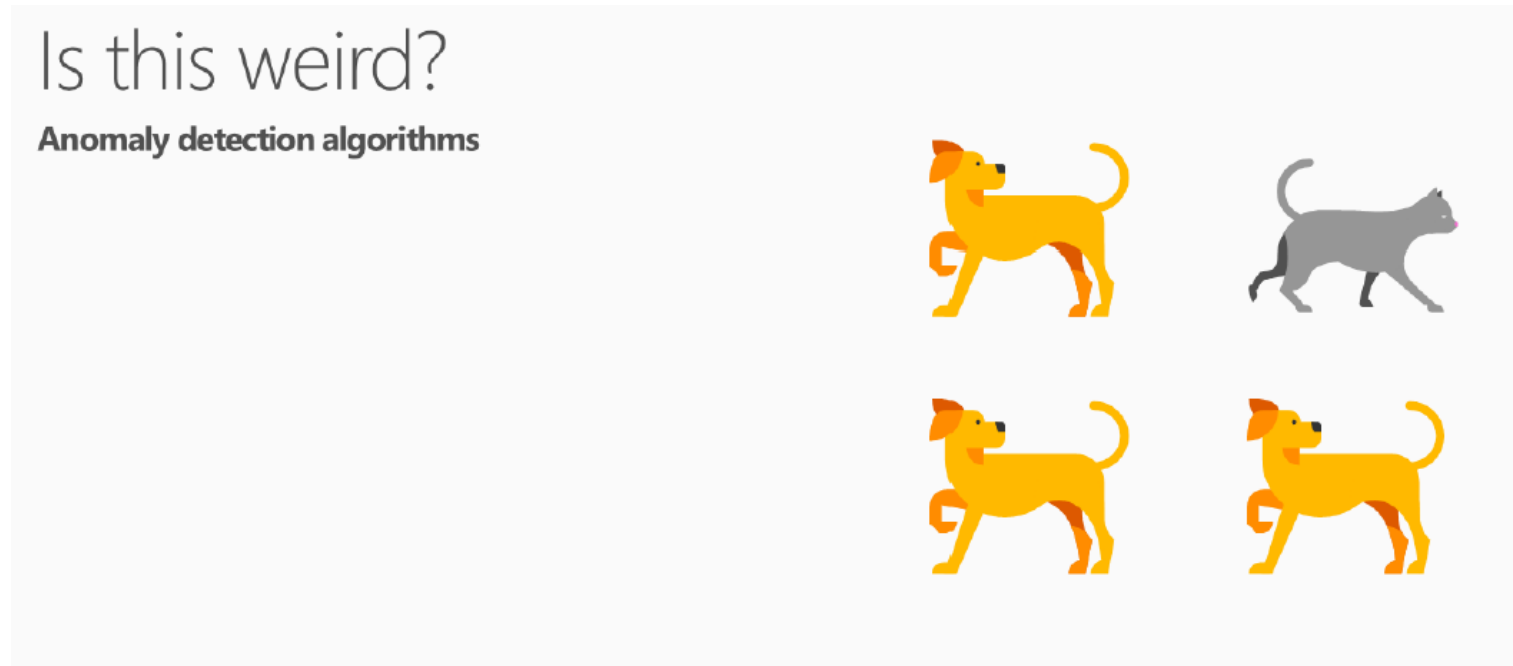
Is this A or B?

Classification algorithms



- Will this tire fail in the next 1,000 miles: Yes or no?
- Which brings in more customers: a \$5 coupon or a 25% discount?
- Can also be more than two options: Is this A or B or C or D, etc.?
- Classification algorithms: helps choosing the most likely one.

## The 5 questions data science answers

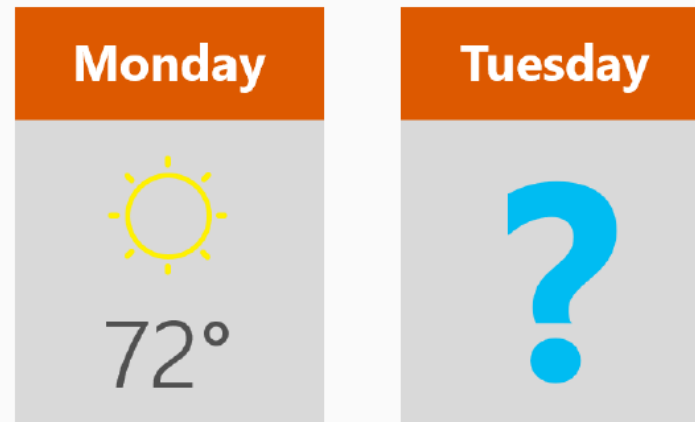


- Is this pressure gauge reading normal?
- Is this message from the internet typical?
- Credit card purchase pattern normal?
- Anomaly algorithms: Detect unexpected or unusual events or behaviors

The 5 questions data science answers

# How much? How many?

Regression algorithms



- What will the temperature be next Tuesday?
- What will my fourth quarter sales be?
- Regression algorithms: Good for question that asks for a number

The 5 questions data science answers

## How is this organized?

### Clustering Algorithms



- Which viewers like the same types of movies?
- Which printer models fail the same way?
- Clustering algorithms: helps arranging data into groups
- Understanding how data is organized, helps predict behaviors and events.

The 5 questions data science answers

## What should I do now?

### Reinforcement Learning Algorithms

- Adjust the temperature or leave it where it is?
- At a yellow light, brake or accelerate?
- Keep vacuuming, or go back to the charging station?
- Reinforcement learning algorithms: the brains of rats and humans respond to punishment and rewards. learning from trial and error.

Is your data ready for data science

We need data that is:

- Relevant
- Connected
- Accurate
- Enough to work with

Is your data ready for data science

## Relevant

Irrelevant Data

Price of milk (\$/gal)	Red Sox batting avg.	Blood alcohol content (%)
3.79	.304	.03
3.45	.320	.09
4.06	.259	.01
3.89	.298	.05
4.12	.332	.13
3.92	.270	.06
3.23	.294	.10

Relevant Data

Body mass (kg)	Margaritas	Blood alcohol content (%)
103	3	.03
67	5	.09
87	1	.01
52	2	.05
73	5	.13
79	3	.06
110	7	.10

- We need to know Blood alcohol content %
- Price of milk and Red Sox are irrelevant

Is your data ready for data science

## Connected

Disconnected Data

Grill temp. (Fahrenheit)	Weight of beef patty (lb)	Burger rating (out of 10)
	.33	8.2
	.24	5.6
550		7.8
725	.45	9.4
600		8.2
625		6.8
	.49	4.2

Connected Data

Grill temp. (Fahrenheit)	Weight of beef patty (lb)	Burger rating (out of 10)
575	.33	8.2
550	.24	5.6
550	.69	7.8
725	.45	9.4
600	.57	8.2
625	.36	6.8
550	.49	4.2

- Quality of hamburgers
- But notice the gaps in the table on the left
- It's common to have holes like this



Is your data ready for data science

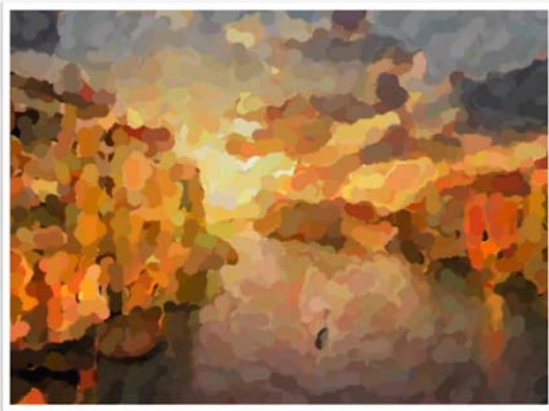
## Accurate



- Top left: precise=yes/accurate=no
- Bottom left: precise=no/accurate=no
- Top right: precise=yes/accurate=yes
- Bottom right: precise=no/accurate=yes

Is your data ready for data science

Enough to work with



## We need enough data to work with

1. Not enough data: can not make decision
2. Barely enough data: can make basic decision ( Is it somewhere I might want to visit? It looks bright, that looks like clean water – yes, that's where I'm going on vacation.)
3. Enough data: can make detailed decision (Now I can look at the three hotels on the left bank. You know, I really like the architectural features of the one in the foreground. I'll stay there, on the third floor.)

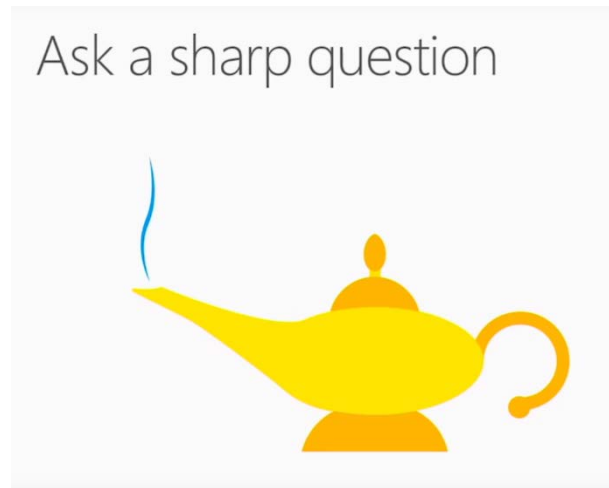
Ask a question you can answer with data

## Ask a question you can answer with data

- Sharp question is the Key
- Know Target Data
- Reword the question

Ask a question you can answer with data

## Sharp question is the Key



- Asking a sharp question is the most important
- ML is a mischievous genie
- "What's going to happen with my stock?", the genie might answer, "The price will change"
- "What will my stock's sale price be next week?", the genie can't help but give you a specific answer and predict a sale price

Ask a question you can answer with data

## Know Target Data

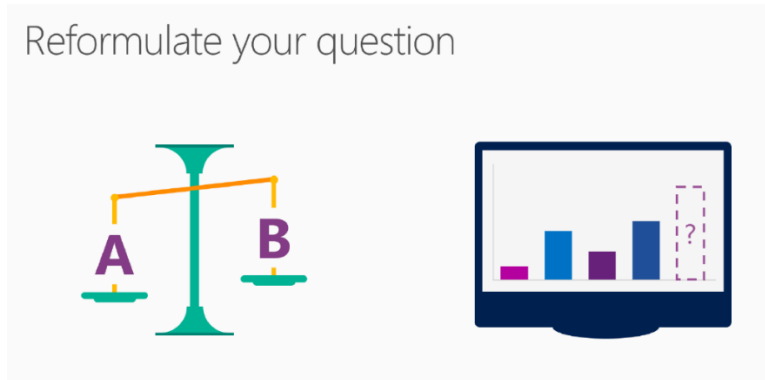
Examples of the answer: Target data



- Target data = what we are trying to predict
- Must have examples of the target in data.
- Question = "What will my stock's sale price be next week?" Target = stock price history.
- Question = "Which car in my fleet is going to fail first?" Target = previous failures data.

Ask a question you can answer with data

## Reword the question



- Question dictates the algorithm
- "Is this data point A or B?" = classification
- "How much?" or "How many?" = regression
- "Which news story is the most interesting to this reader?"
- Algorithm = classification A or B or C or D; difficult
- Reword = "How interesting is each story on this list to this reader?"
- Give each article a numerical score
- Identify the highest-scoring article; easy
- Above example change classification question into a regression question  
Predict an answer with a simple model

## Predict an answer with a simple model

- Collect relevant, accurate, connected, enough data
- Ask a sharp question
- Plot the existing data
- Draw the model through the data points
- Use the model to find the answer
- Create a confidence interval

Predict an answer with a simple model

Collect relevant, accurate, connected, enough data

- I want to know how much 1.35 carat diamond will cost
- Go to jewelry store
- Write down the price of all of the diamonds
- List has two columns
- Each column has a different attribute
- Weight in carats and price
- Each row is a single data point
- Data that represents a single diamond.
- This is a small data set; a table

<u>Carats</u>	<u>price</u>
1.01	7,366
.49	985
.31	544
1.51	9,140
.37	493
.73	3,011
1.53	11,413
.56	1,814
.41	876
.74	2,690
.63	1,190
.6	4,172
2.06	11,764
1.1	4,682
1.31	6,171



Predict an answer with a simple model

This data set meets our criteria for quality:

- Relevant: weight is definitely related to price
- Accurate: we double-checked the prices that we write down
- Connected: there are no blank spaces in either of these columns
- Enough data: to answer our question



Predict an answer with a simple model

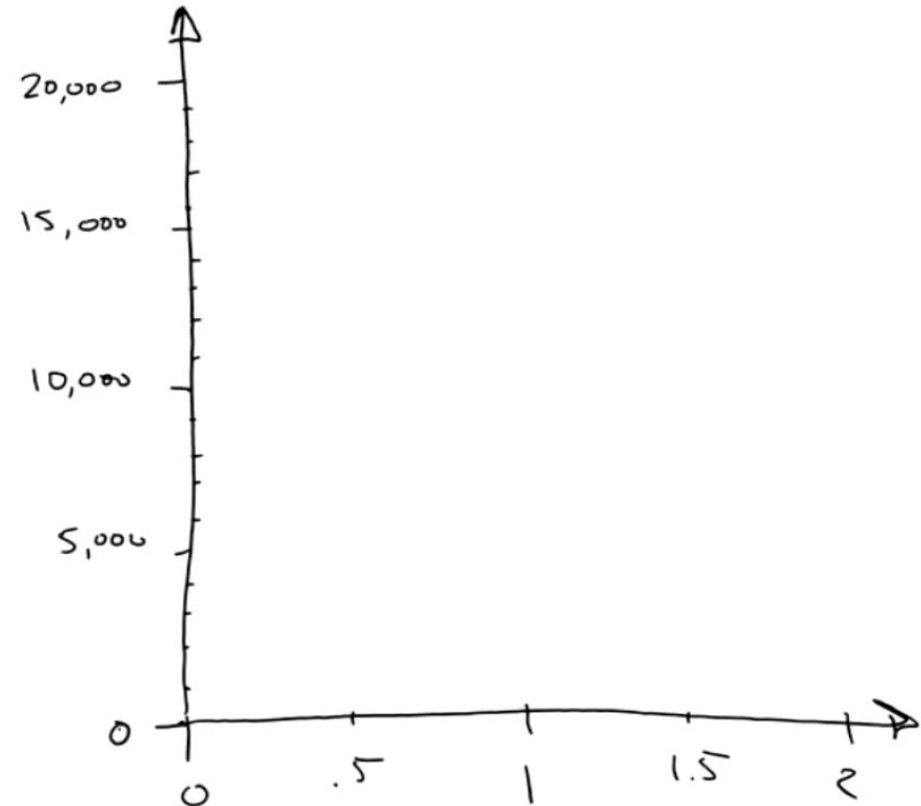
## Ask a sharp question

- How much will it cost to buy a 1.35 carat diamond?
- Our list doesn't have a 1.35 carat diamond
- Use the rest of our data to get an answer to the question

Predict an answer with a simple model

## Draw axis

- Draw a horizontal number line, called an axis, to chart the weights
- The range of the weights is 0 to 2
- Line covers that range and put ticks for each half carat
- Draw a vertical axis to record the price and connect it to the horizontal weight axis
- This will be in units of dollars
- Now we have a set of coordinate axes.

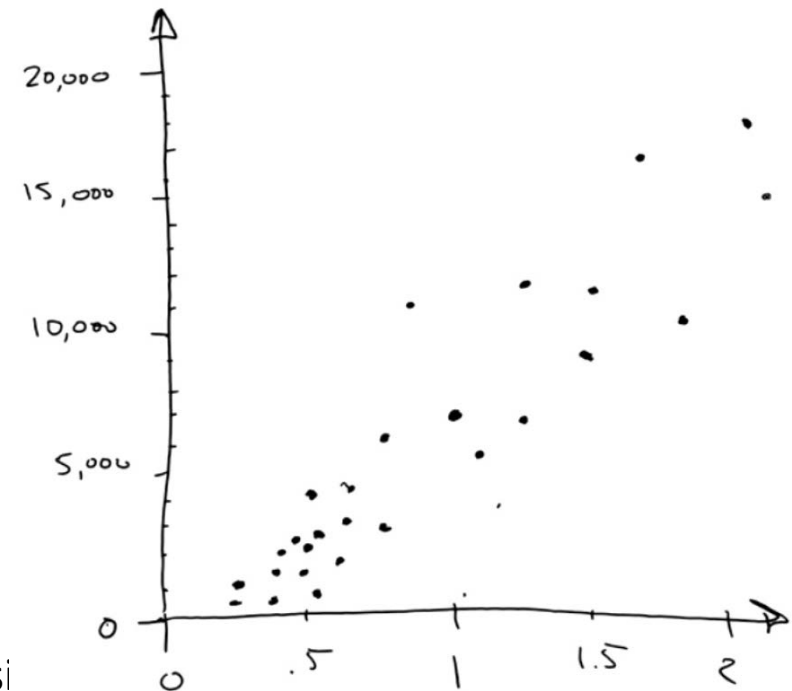


Predict an answer with a simple model

## Plot the existing data

<u>Carats</u>	<u>price</u>
1.01	7,366
.49	985
.31	544

- Make a scatter plot
- Great way to visualize numerical data sets
- For the first data point, we eyeball a vertical line at 1.01 carats. Then, we eyeball a horizontal line at \$7,366. Where they meet, we draw a dot
- This represents our first diamond.
- Now we go through each diamond on this list and do the same thing.
- We get a bunch of dots, one for each diamond



Predict an answer with a si

Draw the model through the data points



- Look at the dots and squint, the collection looks like a fat, fuzzy line
- Draw a straight line through it
- This a model
- Model = cartoon
- The cartoon is wrong
- But, it's a useful simplification
- The line doesn't go through all the data points.
- It has some noise or variance
- But, it's a useful simplification
- Question = How much? regression
- we're using a straight line, linear regression

Predict an answer with a simple model



Use the model to find the answer

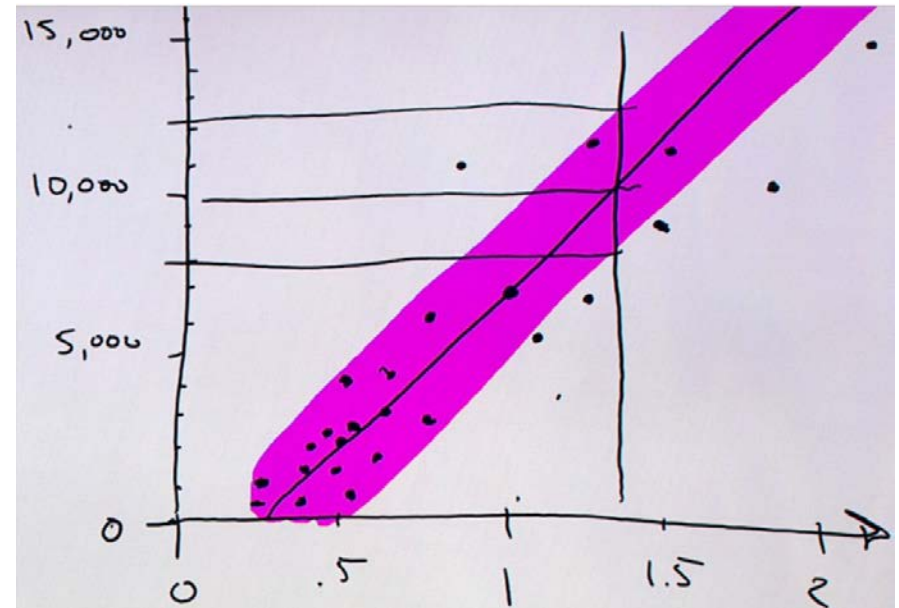


- How much will a 1.35 carat diamond cost?
- Look at 1.35 carats
- Draw a vertical line
- Draw a horizontal line to the dollar axis
- It hits right at 10,000
- Answer = about \$10,000

Predict an answer with a simple model

Create a confidence interval

- How precise this prediction is?
- Is it a lot higher or lower?
- Draw an envelope around the regression line that includes most of the dots.
- This envelope is called our confidence interval
- We're pretty confident that prices fall within this envelope, because in the past most of them have.
- We can draw two more horizontal lines from where the 1.35 carat line crosses the top and the bottom of that envelope.
- The price of a 1.35 carat diamond is about \$10,000 - but it might be as low as \$8,000 and it might be as high as \$12,000



Predict an answer with a simple model

## **We're done, with no math or computers!!**

- We did what data scientists get paid to do, and we did it just by drawing:
- We asked a question that we could answer with data
- We built a model using linear regression
- We made a prediction, complete with a confidence interval

And we didn't use math or computers to do it.





Predict an answer with a simple model

## Now if we'd had more information, like...

- the cut of the diamond
- color variations (how close the diamond is to being white)
- the number of inclusions in the diamond

...then we would have had more columns. In that case, math becomes helpful. If you have more than two columns, it's hard to draw dots on paper. The math lets you fit that line or that plane to your data very nicely.

Also, if instead of just a handful of diamonds, we had two thousand or two million, then you can do that work much faster with a computer.

$$a = \frac{n \sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{n \sum_{i=1}^n x_i^2 - \left( \sum_{i=1}^n x_i \right)^2}$$

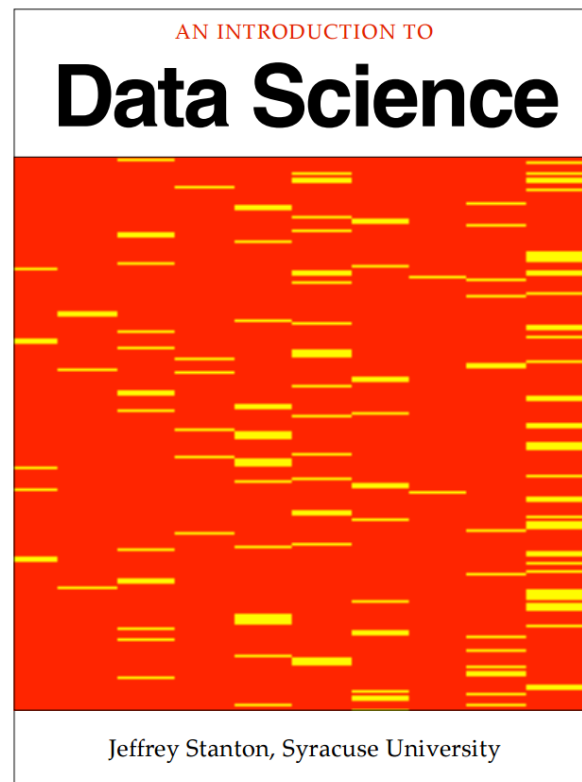
$$b = \frac{1}{n} \left( \sum_{i=1}^n y_i - a \sum_{i=1}^n x_i \right)$$

More information

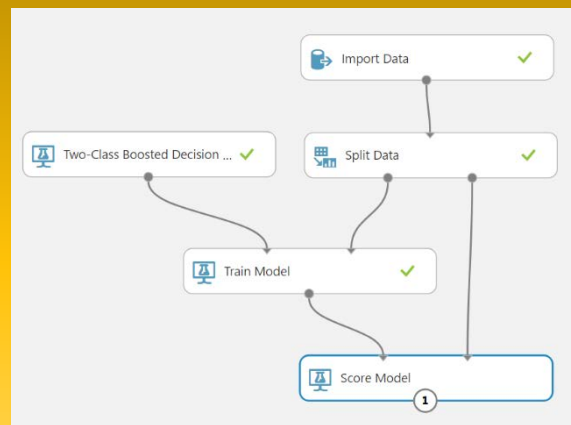
## More information on Data Science Basic

An introduction to Data Science: Jeffrey Stanton

[https://ischool.syr.edu/media/documents/2012/3/DataScienceBook1\\_1.pdf](https://ischool.syr.edu/media/documents/2012/3/DataScienceBook1_1.pdf)



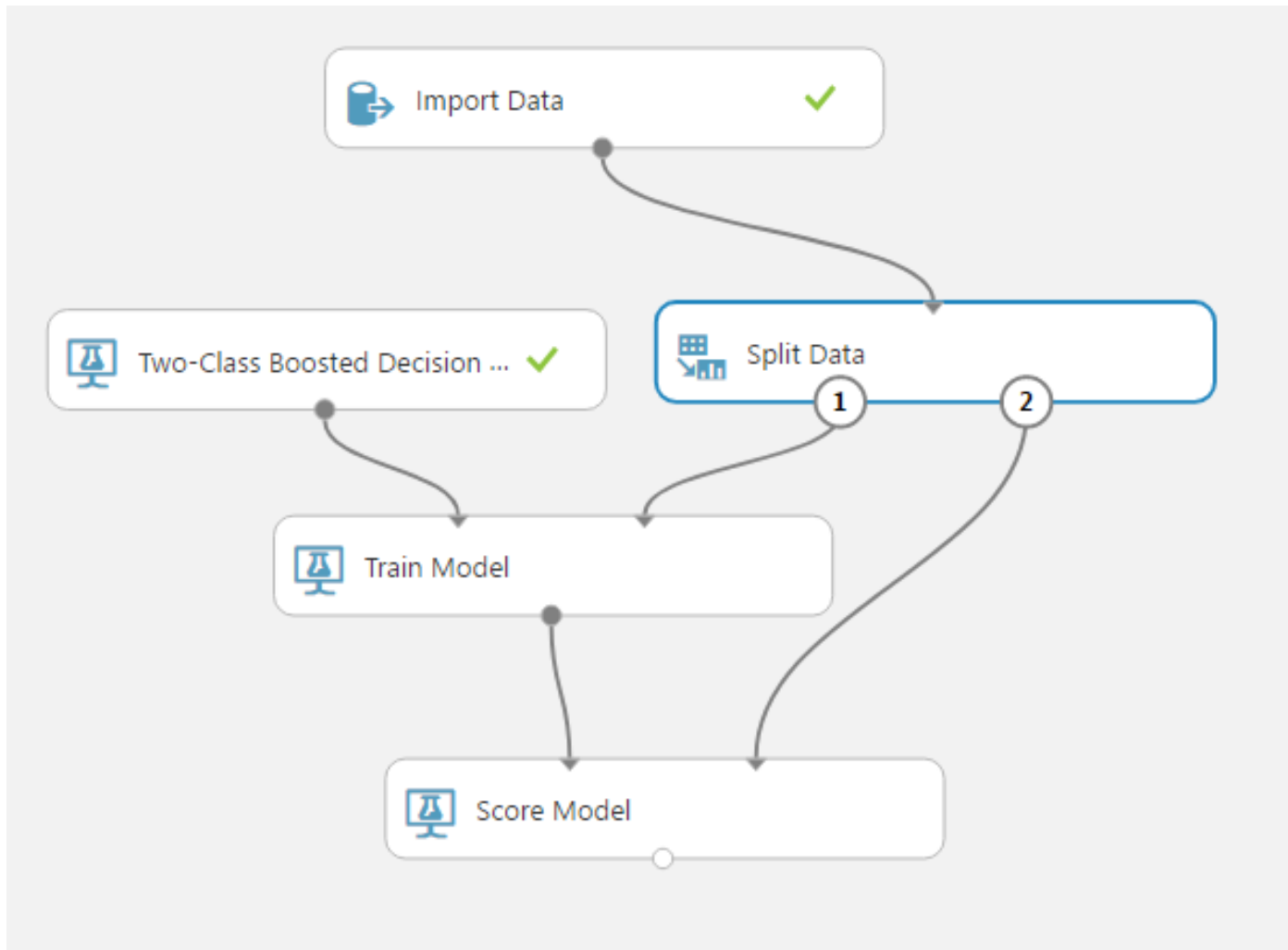
# FIRST EXPERIMENT



## In this session

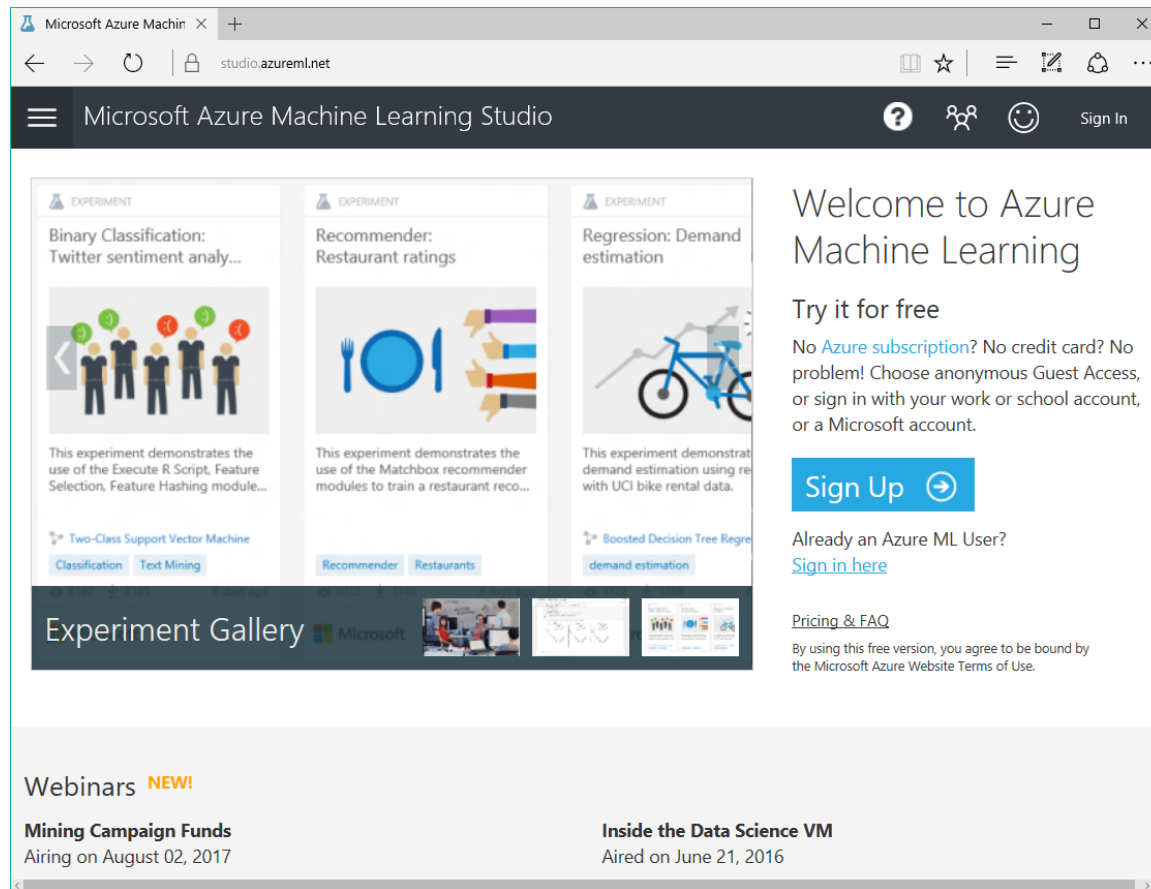
- Sing Up FREE Azure ML Studio Subscription
- Create Azure ML Studio workspace
- Train, Test, Evaluate for Binary Classification
- Import census income dataset
- Create a new Azure Machine Learning experiment
- Train and evaluate a prediction model
- Type of datasets

## First experiment model



## Sing Up FREE Azure ML Studio Subscription

<https://studio.azureml.net/>



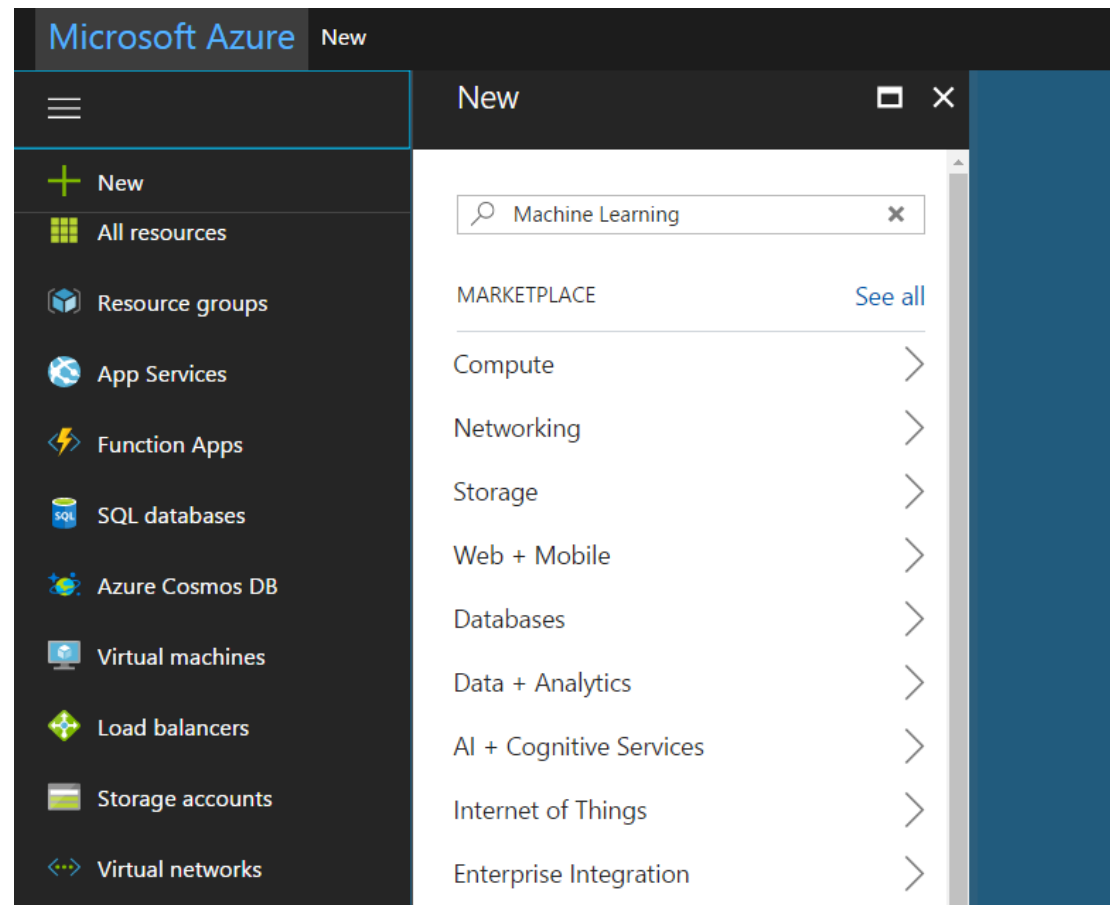
## Sing Up FREE Azure ML Studio Subscription

Free Workspace -> sign up here

Quick Evaluation	Most Popular	Enterprise Grade
Guest Workspace	Free Workspace	Standard Workspace
8-hour trial	\$0/month	\$9.99/month
No sign-in required.	Don't already have a Microsoft account? Simply <a href="#">sign up here</a> .	<a href="#">Azure subscription</a> required Other charges may apply. <a href="#">Read more</a> .
<a href="#">Enter</a>	<a href="#">Sign In</a>	<a href="#">Create Workspace</a>
<ul style="list-style-type: none"><li>▪ No hassle instant access</li><li>▪ Stock sample datasets</li><li>▪ ML models built in minutes</li><li>▪ Full range of ML algorithms</li></ul>	<ul style="list-style-type: none"><li>▪ Free access that never expires</li><li>▪ 10 GB storage on us</li><li>▪ R and Python scripts support</li><li>▪ Predictive web services</li></ul>	<ul style="list-style-type: none"><li>▪ Full SLA Support</li><li>▪ Bring your own Azure storage</li><li>▪ Parallel graph execution</li><li>▪ Elastic Web Service endpoints</li></ul>

## Create Azure ML Studio workspace

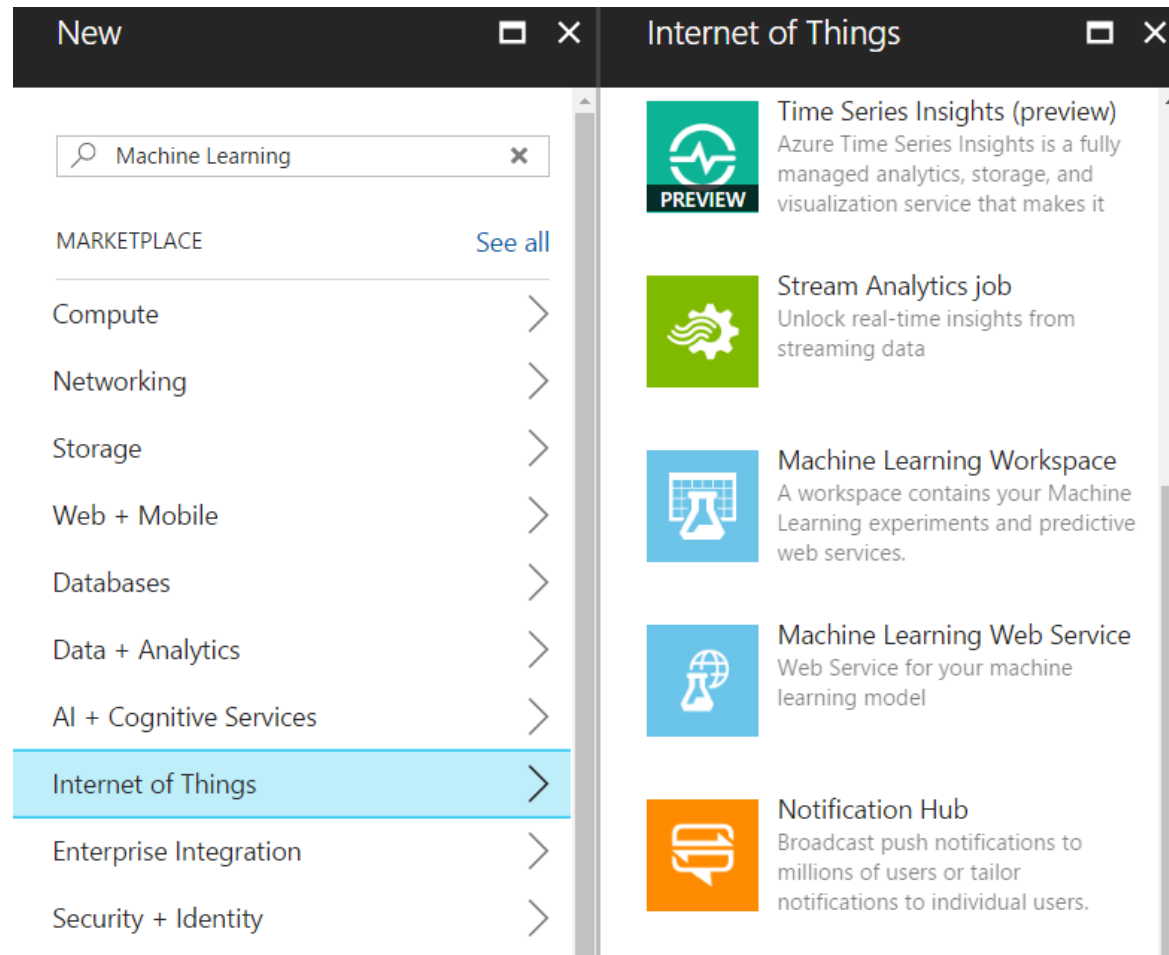
1. Go to the Azure portal <https://portal.azure.com>
2. Click +New





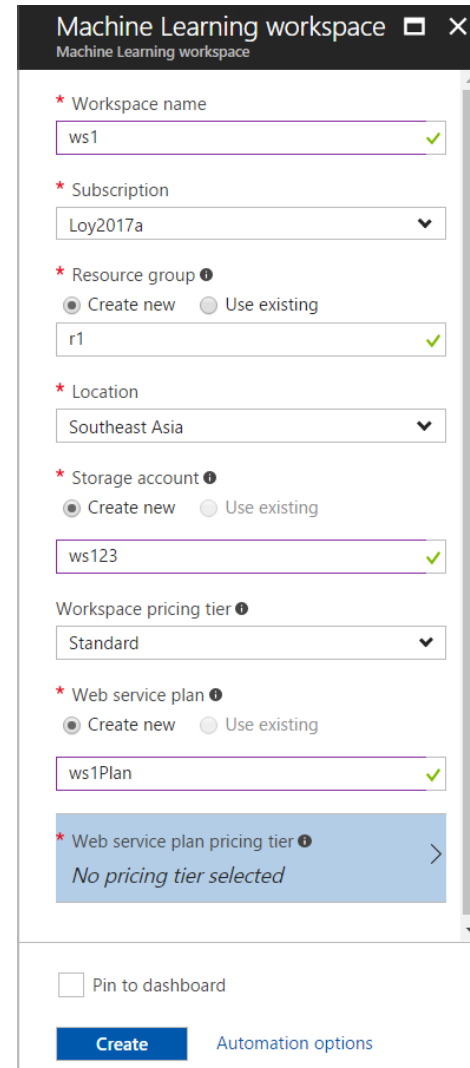
## Create Azure ML Studio workspace

### 3. Select Internet of Things, click Machine Learning Workspace, then click Create



## Create Azure ML Studio workspace

4. Workspace name = ws1
5. Subscription = default
6. Resource group = Create new: rs1
7. Location = Southeast Asia
8. Storage account = Create new: names1
9. Workspace pricing tier = Standard
10. Web service plan = Create new: ws1Plan



The screenshot shows the 'Machine Learning workspace' creation form in the Azure portal. The form is titled 'Machine Learning workspace' with a close button. It contains several fields and options:

- Workspace name:** A text input field with 'ws1' entered and a green checkmark.
- Subscription:** A dropdown menu showing 'Loy2017a'.
- Resource group:** Radio buttons for 'Create new' (selected) and 'Use existing'. Below is a text input field with 'r1' and a green checkmark.
- Location:** A dropdown menu showing 'Southeast Asia'.
- Storage account:** Radio buttons for 'Create new' (selected) and 'Use existing'. Below is a text input field with 'ws123' and a green checkmark.
- Workspace pricing tier:** A dropdown menu showing 'Standard'.
- Web service plan:** Radio buttons for 'Create new' (selected) and 'Use existing'. Below is a text input field with 'ws1Plan' and a green checkmark.
- Web service plan pricing tier:** A blue box with the text 'No pricing tier selected' and a right arrow.

At the bottom, there is a checkbox for 'Pin to dashboard' and a blue 'Create' button. To the right of the 'Create' button is a link for 'Automation options'.

## Create Azure ML Studio workspace

11. Click No pricing tier selected
12. Click DEVTEST
13. Click Pin to dashboard
14. Click Create

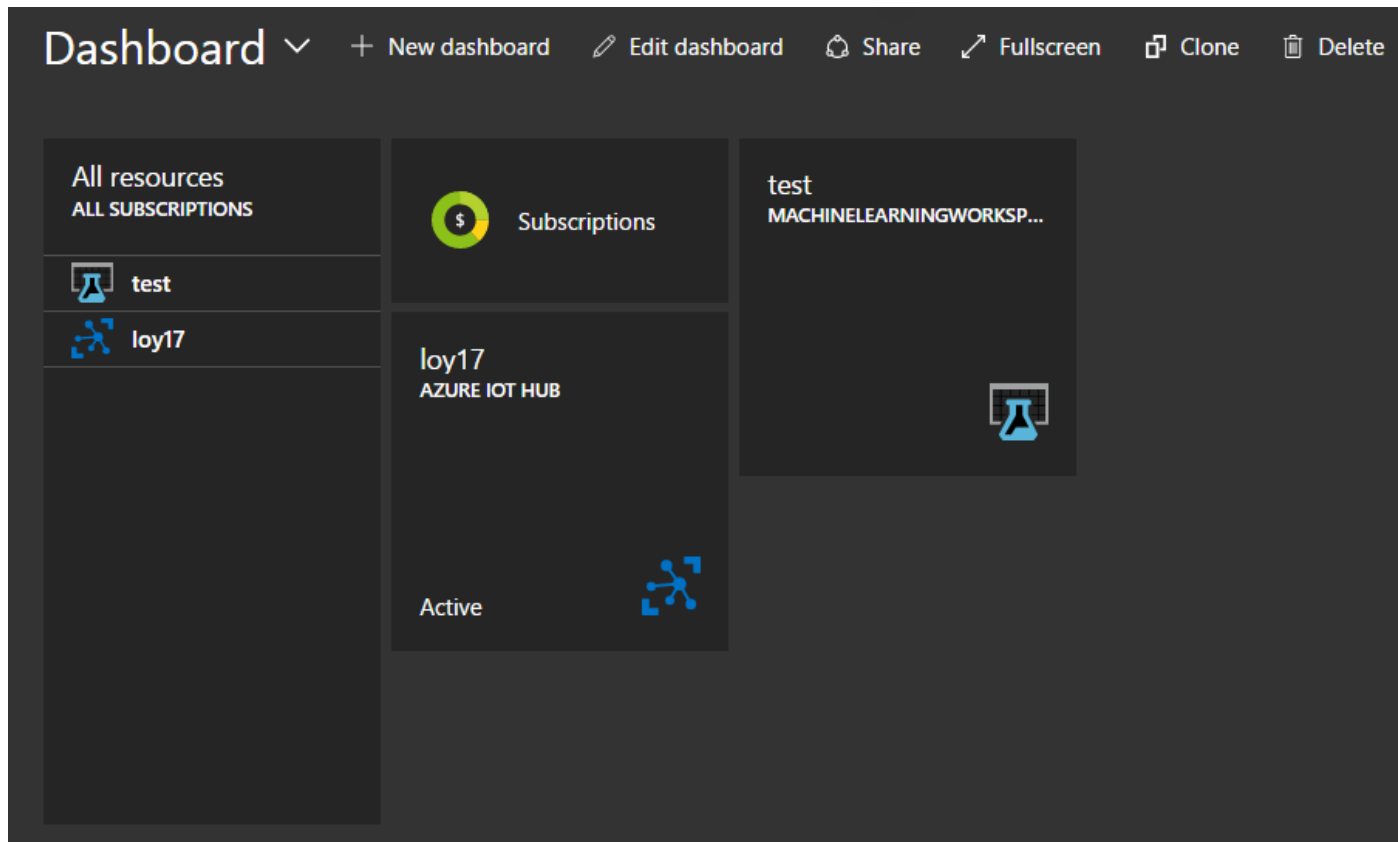
**Choose your pricing tier**  
Browse the available plans

Choose the pricing tier for Machine Learning web services. [Learn more](#)

DEVTEST Standard	S1 Standard	S2 Standard
2 Compute Hours	25 Compute Hours	500 Compute Hours
1,000 Transactions	100,000 Transactions	2,000,000 Transactions
Manual Scaling	Manual Scaling	Manual Scaling
0.00 USD/DAY (ESTIMATED)	3.23 USD/DAY (ESTIMATED)	32.26 USD/DAY (ESTIMATED)

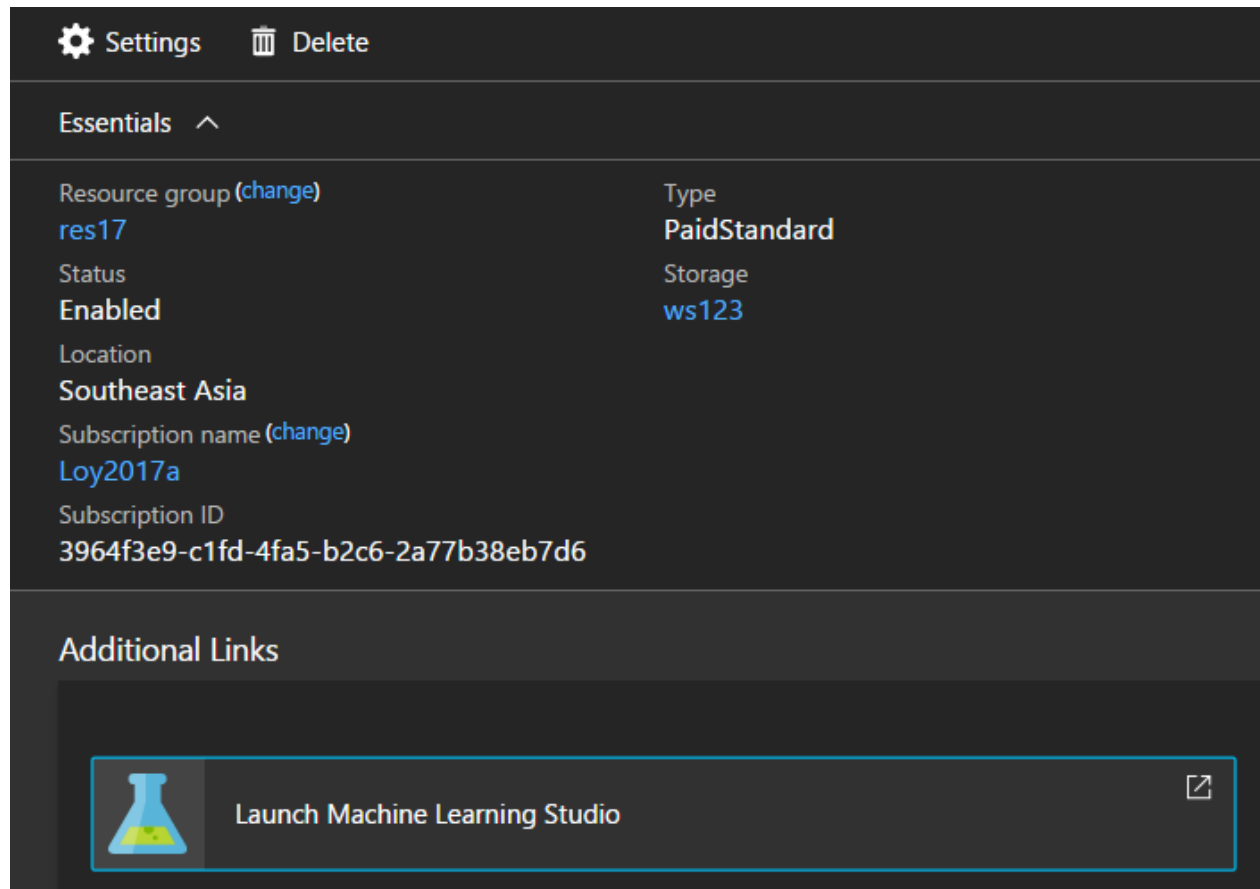
## Create Azure ML Studio workspace

15. Click at Machine Learning workgroup on dashboard



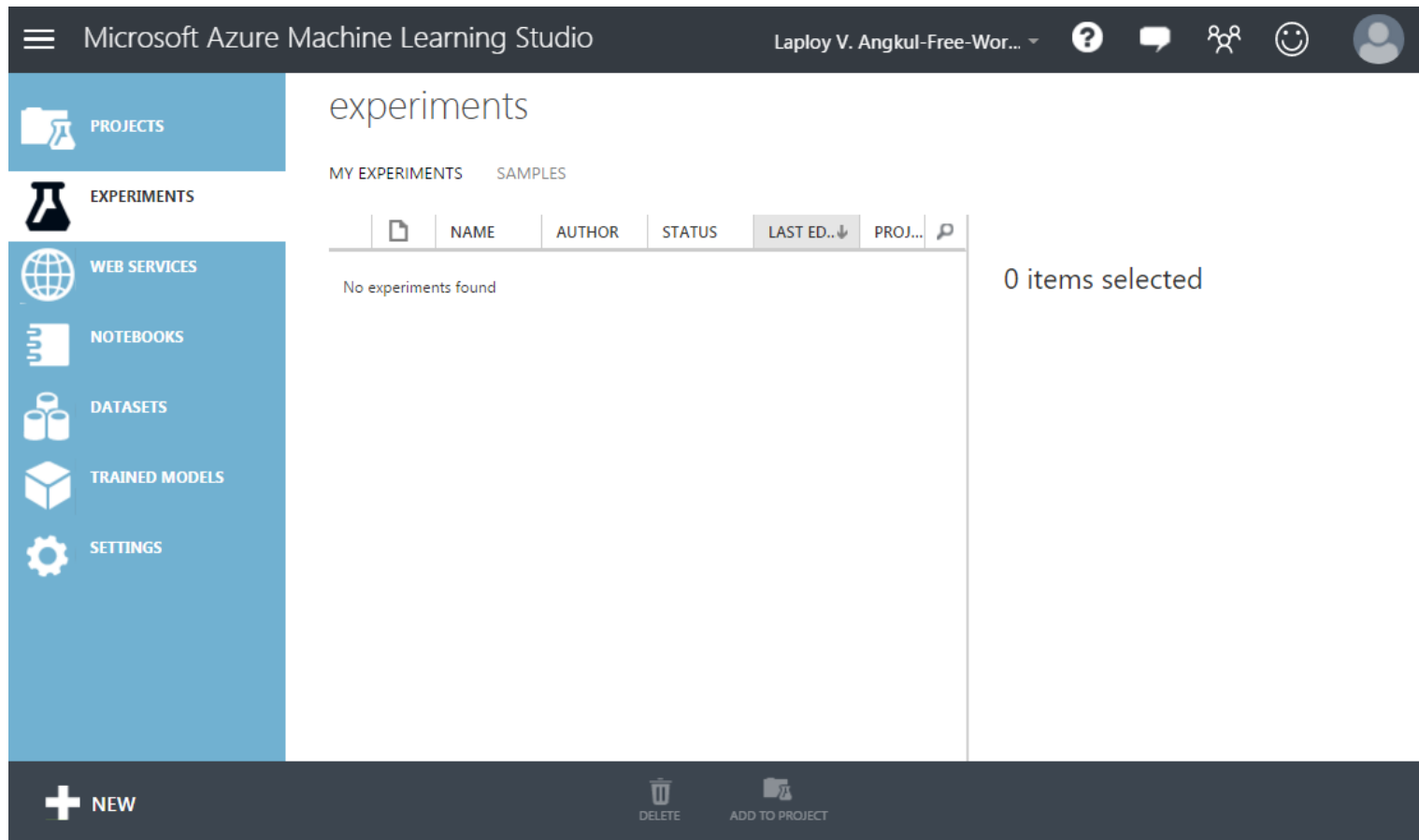
## Create Azure ML Studio workspace

### 16. Click Launch Machine Learning Studio



## Create Azure ML Studio workspace

### Blank, new ML Studio workspace



## Train, Test, Evaluate for Binary Classification

Predicting whether a person's income exceeds \$50,000 per year based on his demographics or census data

1. Download, prepare, and upload a census income dataset.
2. Create a new Azure Machine Learning experiment.
3. Train and evaluate a prediction model.

The overall workflow of the experiment

Microsoft Azure Machine Learning Studio

Laploy V. Angkul-Free-Wor... ?

Training experiment Predictive experiment

Adult Income 2 In draft

Draft saved at 9:31:10 PM

Search experiment items

- Saved Datasets
- Trained Models
- Transforms
- Custom
- Data Format Conversions
- Data Input and Output
- Data Transformation
- Feature Selection
- Machine Learning
- OpenCV Library Modules
- Python Language Modules
- R Language Modules
- Statistical Functions
- Text Analytics
- Time Series
- Web Service
- Deprecated

Import Data

Two-Class Boosted Decision ...

Split Data

Train Model

Score Model

Splitting mode

Split Rows

Fraction of rows in the first...

0.80

☒ Randomized split

Random seed

0

Stratified split

False

Quick Help

Split the rows of a dataset into two distinct sets  
(more help...)

+ NEW

RUN HISTORY

SAVE

SAVE AS

DISCARD CHANGES

RUN

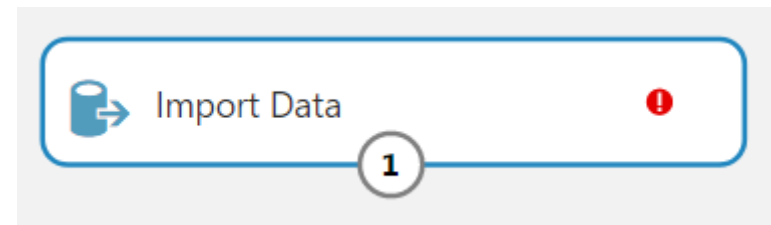
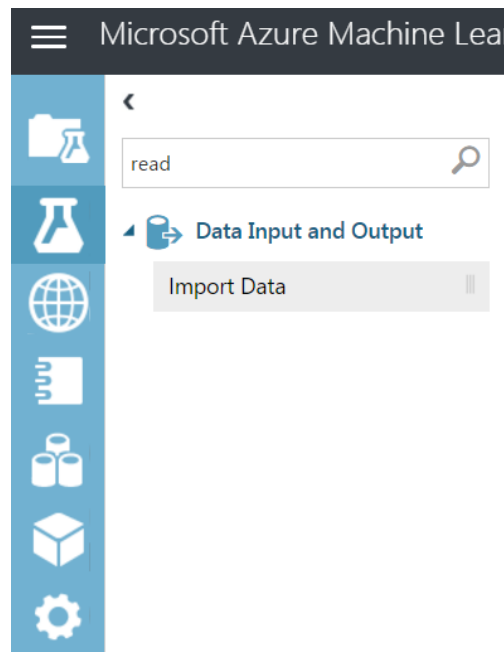
SET UP WEB SERVICE

PUBLISH TO GALLERY



## Train, Test, Evaluate for Binary Classification

- Create New blank experiment. Name = Adult Income 1
- Click Data Input and Output
- Drag & drop Import Data



## Train, Test, Evaluate for Binary Classification

Configure Import data module:

- Data source = Web URL via HTTP
- Data source URL = <http://archive.ics.uci.edu/ml/machine-learning-databases/adult/adult.data>
- Data format = CSV
- Run experiment

Properties Project

### Import Data

Launch Import Data Wizard

Data source

Web URL via HTTP

Data source URL

<http://archive.ics.uci.edu/ml/machine-learning-databases/adult/adult.data>

Data format

CSV

☐ CSV or TSV has header row

☐ Use cached results

START TIME 5/20/2017 9:13:05 PM

END TIME 5/20/2017 9:13:16 PM

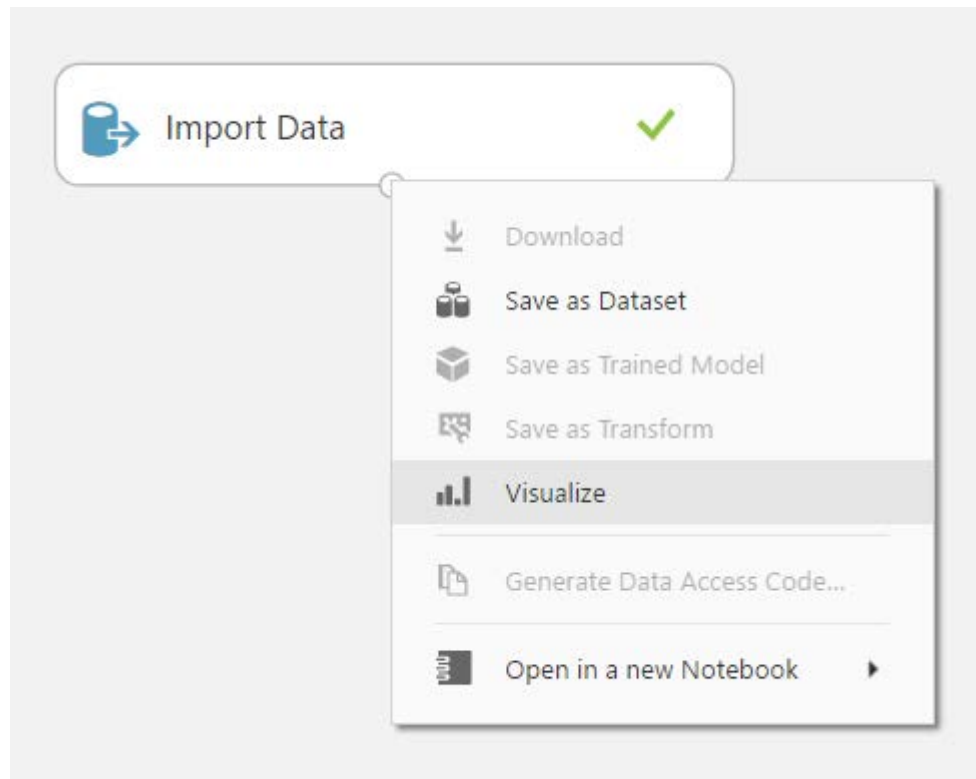
ELAPSED TIME 0:00:11.502

STATUS CODE Finished

STATUS DETAILS None

## Train, Test, Evaluate for Binary Classification

- Right click at the output of Import Data
- Click Visualize









## Train, Test, Evaluate for Binary Classification

- Click on Col2
- Look at Statistics and Histogram

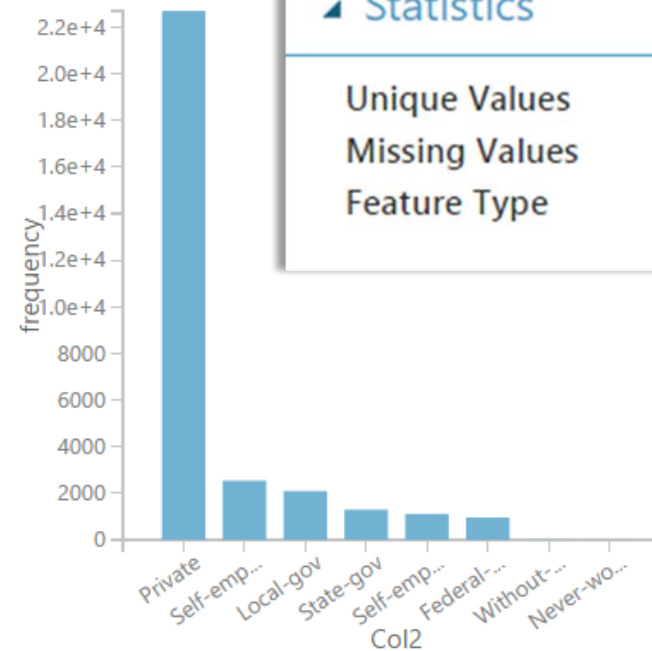
Adult Income > Import Data > Results dataset

rows  
32562

columns  
15

	Col1	Col2	Col3	Col4	Col5	Col6
view as						
	39	State-gov	77516	Bachelors	13	Never-married
	50	Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse
	38	Private	215646	HS-grad	9	Divorced
	53	Private	234721	11th	7	Married-civ-spouse
	28	Private	338409	Bachelors	13	Married-civ-spouse
	37	Private	284582	Masters	14	Married-civ-spouse

Col2  
Histogram



### Statistics

Unique Values	8
Missing Values	1837
Feature Type	String Feature

## Train, Test, Evaluate for Binary Classification

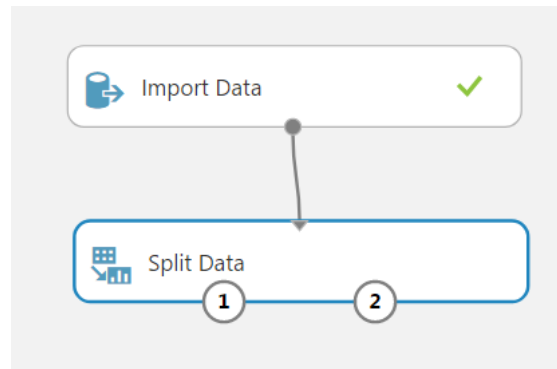
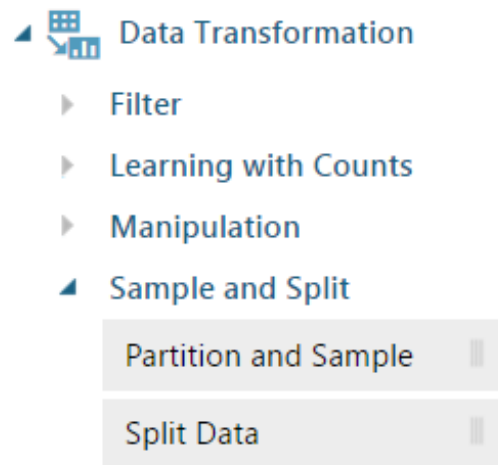
### Split up the dataset

- Training data: This grouping is used for creating our new predictive model based on the inherent patterns found in the historical data via the ML algorithm we use for the solution.
- Validation data: This grouping is used for testing the new predictive model against known outcomes to determine accuracy and probabilities.

## Train, Test, Evaluate for Binary Classification

## Add Split Data:

- Click Data Transformation
- Click Sample and Split
- Drag & drop Split Data module into canvas
- Connect Import Data to Split Data
- Set properties Fraction of row to 0.80



## Properties Project

### Split Data

Splitting mode

Split Rows ▼

Fraction of rows in the first... ⋮

0.80

☒ Randomized split ⋮

Random seed ⋮

0

Stratified split

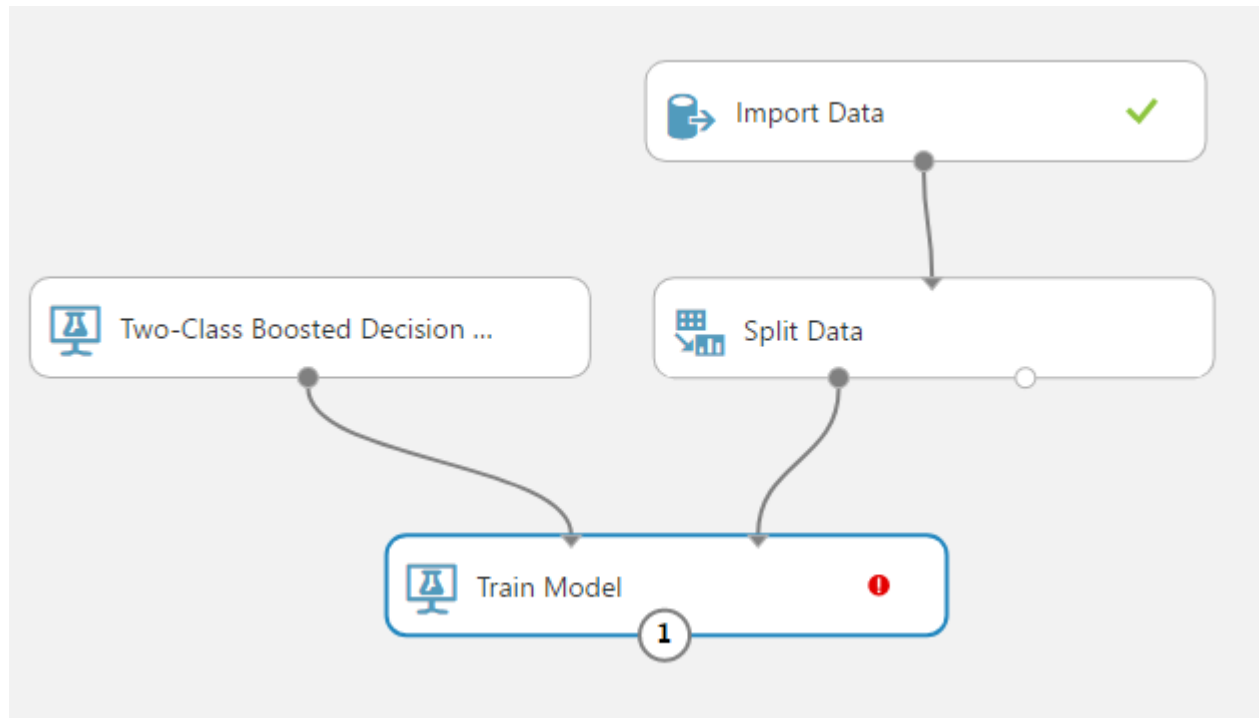
False ▼

## Train, Test, Evaluate for Binary Classification

Add Two-Class Boosted Decision Tree and Train Model

Connect Two-Class Boosted Decision Tree to Train Model

Connect Split Data to Train Model



### Two-Class Boosted Decision T..

Create trainer mode

Single Parameter ▼

Maximum number of le... ⋮

20

Minimum number of sa... ⋮

10

Learning rate ⋮

0.2

Number of trees constru.. ⋮

100

Random number seed ⋮

☒ Allow unknown cate... ⋮

## Train, Test, Evaluate for Binary Classification

Click Train Model

Click Launch column selector

Include col15

Click 

Save

Run

Select a single column

BY NAME

WITH RULES

Include ▼ column names ▼

Col15 ✕

Score the model

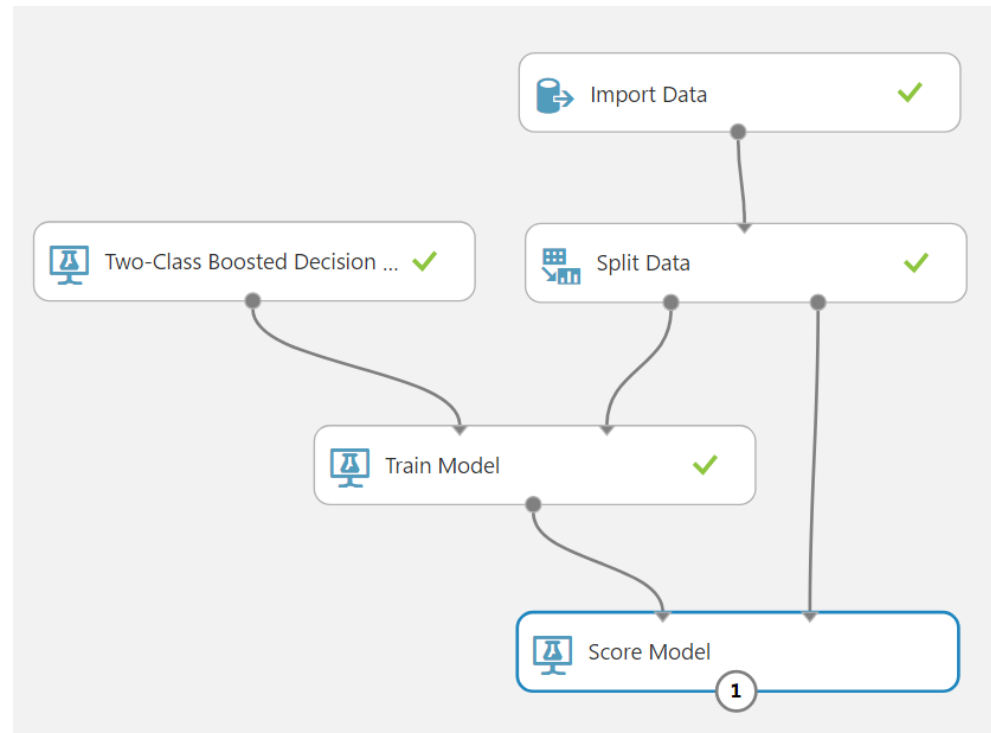


## Score the model:








Add Score Model to canvas

Connect Score Model to Train and Split model

Run



## Visualize the model results

Col11	Col12	Col13	Col14	Col15	Scored Labels	Scored Probabilities
						
0	0	50	United-States	<=50K	<=50K	0.425173
0	0	40	Puerto-Rico	<=50K	<=50K	0.008254
0	0	35	United-States	<=50K	<=50K	0.002206

Visualize the model results:

Visualize output of Score Model

**Scored Labels** This column denotes the model's prediction for this row of the dataset.

**Scored Probabilities** This column denotes the numerical probability (or the likelihood) of whether the income level for this row exceeds \$50,000.

## Type of datasets

### **Training set**

- A set of examples used for learning
- Where the answer value is known.

### **Validation set**

- A set of examples data
- Used to tune the architecture of a classifier
- And estimate the error

### **Test set**

- Use to test the performances of a classifier
- Never used during the training process
- Give estimate of error

More Information