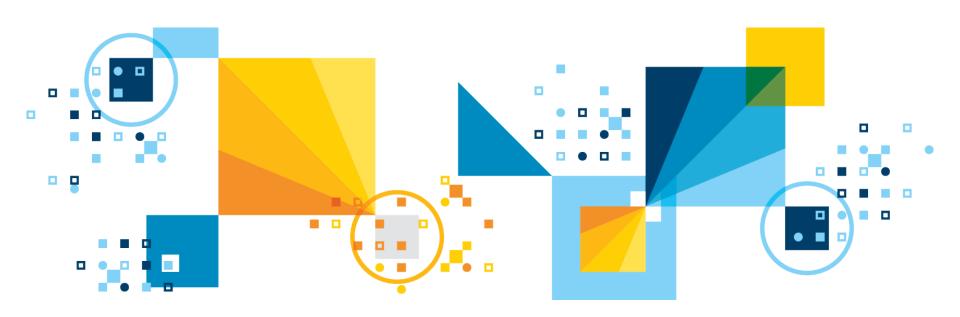
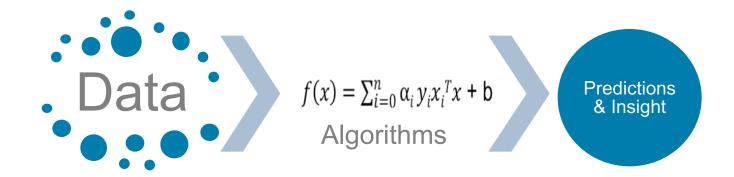
# Introduction to Machine Learning

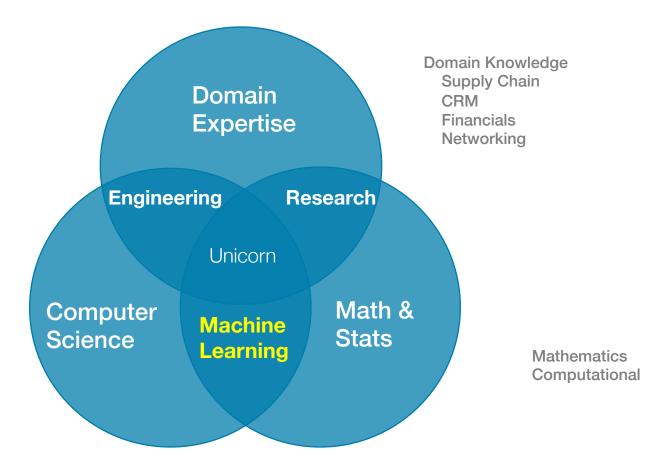


# What is Machine Learning?

"Computers that learn without being explicitly programmed"
"Using algorithms to understand patterns in data"



# Machine Learning in Data Science Echosystem....

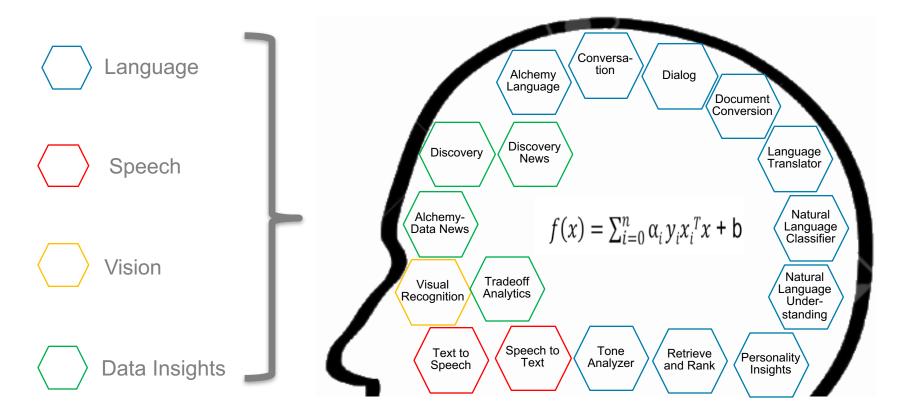


Scripting, SQL Python, R Scala Data Pipelines Big Data/ Apache Spark

Data Science Projects Require multiple Skills

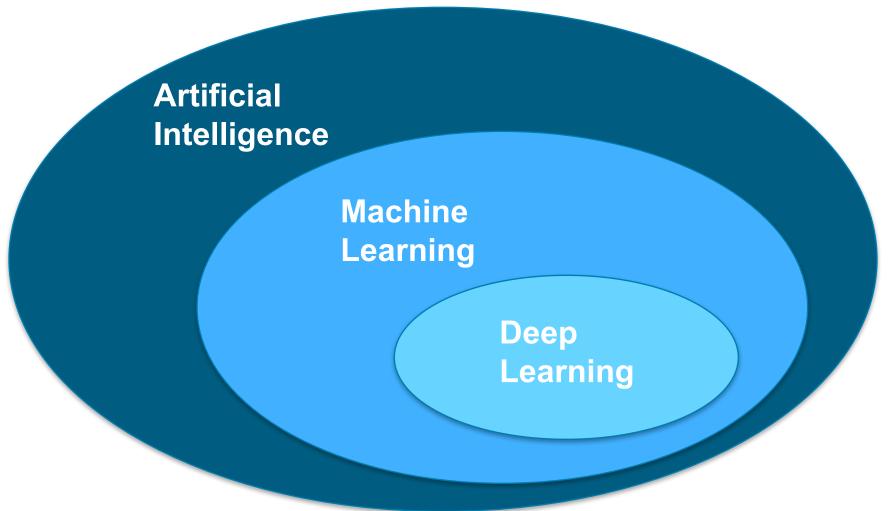
# **Machine Learning = Artificial Intelligence???**

# Data + Algorithms = Scored Al Models



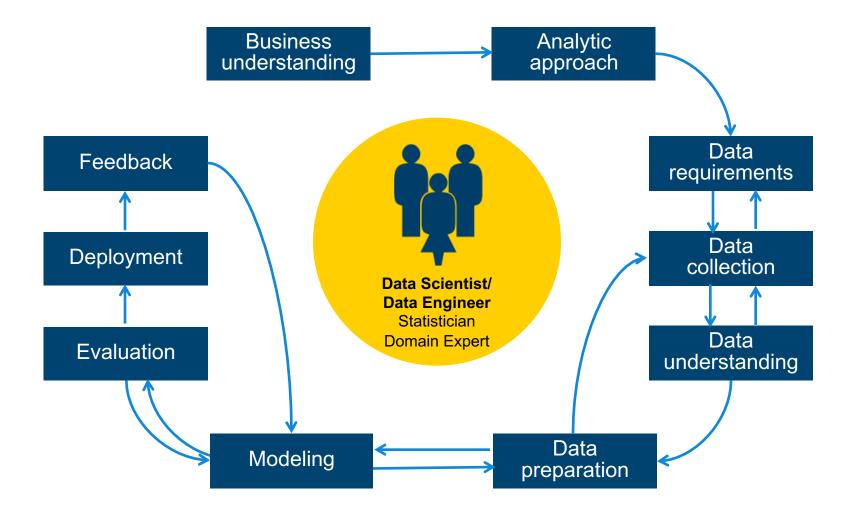


# **Understanding AI, ML & DL Relationship...**



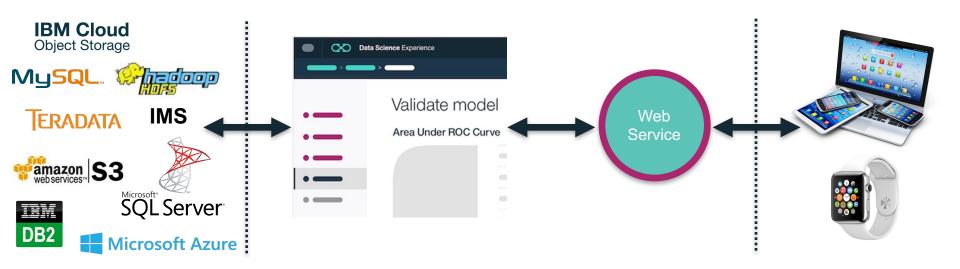


# Methodology to create Machine Learning Model





# Introducing IBM Watson Machine Learning Accelerating Digital Deployments in Cloud...



#### **Data Access:**

- Easily connect to Behind-the-Firewall and Public Cloud Data
- Catalogued and Governed Controls through Watson Data Platform

#### **Creating Models:**

- Single UI and API for creating ML Models on various Runtimes
- Auto-Modelling and Hyperparameter Optimization

#### Web Service:

- Real-time,
   Streaming, and
   Batch Deployment
- Continuous
   Monitoring and
   Feedback Loop

#### **Intelligent Apps:**

- Integrate ML models with apps, websites, etc.
- Continuously Improve and Adapt with Self-Learning

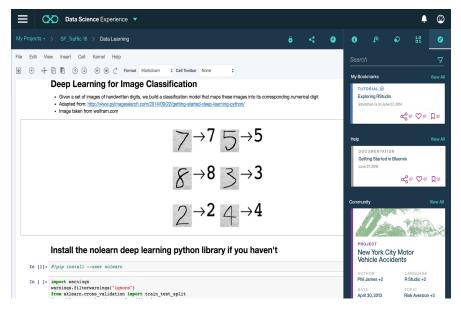


#### IBM Watson Machine Learning in Data Science Experience



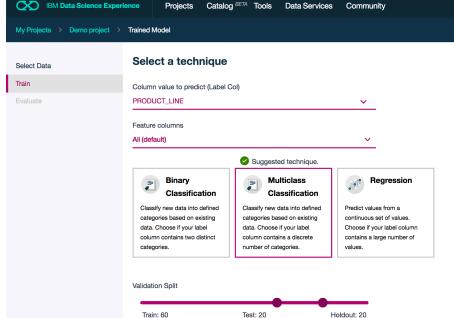
### APIs for Jupyter Notebooks







Open Beta

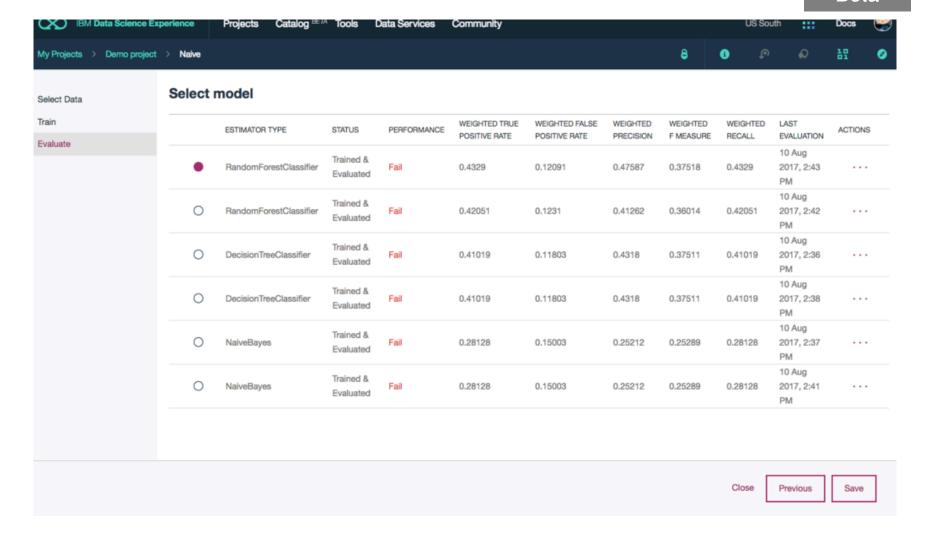




■Data Scientist

Open

# WML GUI - Train multiple models at same ting System tells you best model...



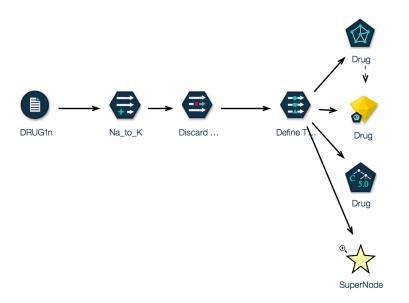


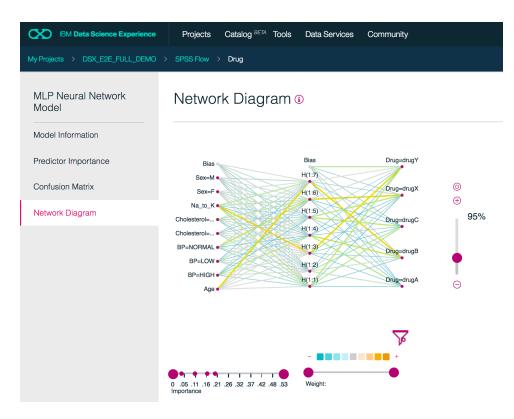
#### **IBM Watson Machine Learning in Data Science Experience**



**Flows** 

Open Beta





Create Advanced Models without coding

Advanced Model Visualization: Easy to understand their performance

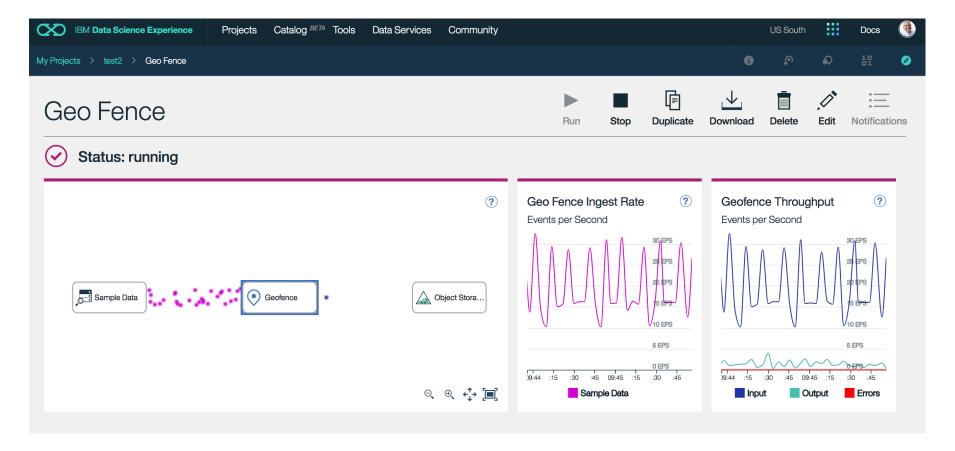


#### **IBM Watson Machine Learning in Data Science Experience**



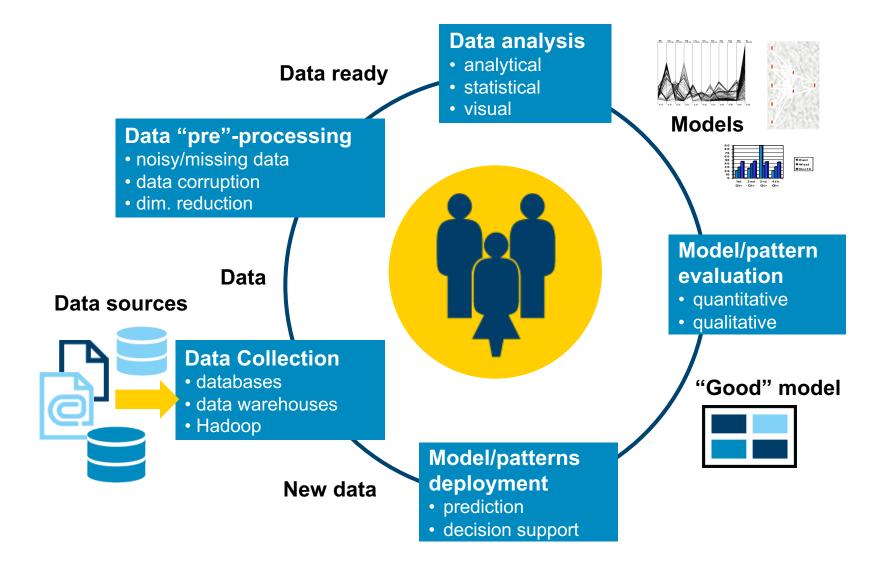
#### **Streaming Pipelines**







# Data centric view of methodology



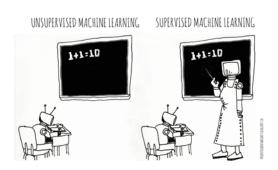
# **Categories of Machine Learning**

#### Supervised learning

- The program is "trained" on a pre-defined set of "training examples", which then facilitate its ability to reach an accurate conclusion when given new data
- The algorithm is presented with example inputs and their desired outputs (correct results)
- The goal is to learn a general rule that maps inputs to outputs

#### Unsupervised learning

- No labels are given to the learning algorithm, leaving it on its own to find structure (patterns and relationships) in its input
- Unsupervised learning can be a goal in itself (discovering hidden patterns in data) or a means towards an end (feature learning)





# **Categories of Machine Learning**

	Discrete Output	Continuous Output
Supervised Learning (require Ground-Truth)	<ul> <li>Classification (outcome is discrete)</li> <li>Binary Classification</li> <li>Linear Models (Logistic Regression)</li> <li>Decision Trees</li> <li>Naïve Bayes</li> <li>Multi class Classification</li> <li>Decision Trees</li> <li>Naïve Bayes</li> <li>K-NN</li> </ul>	<ul> <li>Regression <ul><li>Linear</li><li>Ridge</li><li>Lasso</li></ul> </li> <li>Decision Trees <ul><li>Random Forest</li><li>Gradient Boosted Trees</li></ul> </li> </ul>
Unsupervised Learning (no Ground-Truth data required)	<ul><li>Clustering     - k-means</li><li>FP-Growth</li></ul>	<ul> <li>Clustering <ul><li>k-means</li><li>Gaussian Mixture</li></ul> </li> <li>Dimensionality Reduction <ul><li>PCA</li><li>SVD</li></ul> </li> </ul>

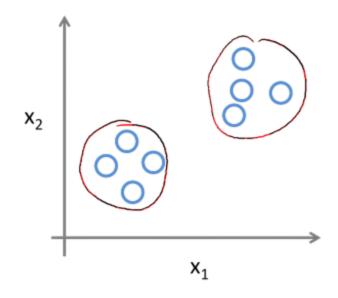


# Supervised vs. Unsupervised Learning

## Supervised Learning

# $x_2$ $x_2$ $x_2$ $x_1$

# **Unsupervised Learning**





Modeling

# Training, testing, & validation sets

Evaluation

- During the model development process, supervised learning techniques employ training and testing sets and sometimes a validation set.
  - Historical data with known outcome (target, class, response, or dependent variable)
  - Source data randomly split or sampled... mutually exclusive records

#### Why?

- Training set → build the model (iterative)
- Testing set → tune the parameters & variables during model building (iterative)
  - Assess model quality during training process
  - Avoid overfitting the model to the training set
- Validation set → estimate accuracy or error rate of model (once)
  - Assess model's expected performance when applied to new data

# **Machine Learning – A more formal definition**

Tom Mitchell of Carnegie Mellon University provides a widely quoted, more formal definition of machine learning

"A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P if its performance at tasks in T, as measured by P, improves with experience E"





**Q & A** 

