# How does Data Science fit into Consulting?

#### **Pharmaceutical Client**

Optimizing how drugs are brought to market

#### **Business Goals**

- Improve clinical operations clinical trial productivity by more than 10%
- Better monitor site level risks

#### **Data Science Goals**

- Country footprint optimization
- site selection and risk modeling
- trial management and forecasting



https://www.mckinsey.com/business-functions/mckinseyanalytics/how-we-help-clients



### Agenda

- Types of Machine Learning
- Classification Regression Problems
- Train Test Split
- Preprocessing
- Logistic Regression
- Loss Function
- Model Evaluation
- Assignment
- Further Reading



# Types of Machine Learning

#### Supervised Learning

$$f(x_i) \approx y_i$$

Examples: price prediction, spam detection, medical diagnosis, ad click prediction

Unsupervised Learning

$$x_i \propto p(x)$$

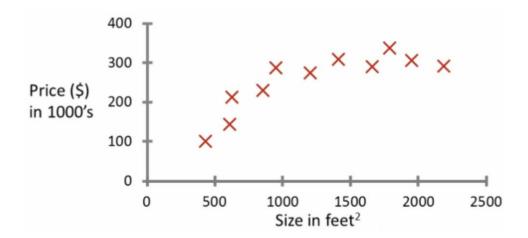
Examples: Clustering, Dimensionality reduction



# Classification and Regression Problems

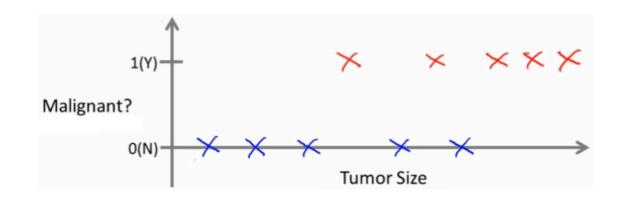
#### Regression

- Target y is continuous
- Example: House price prediction



#### Classification

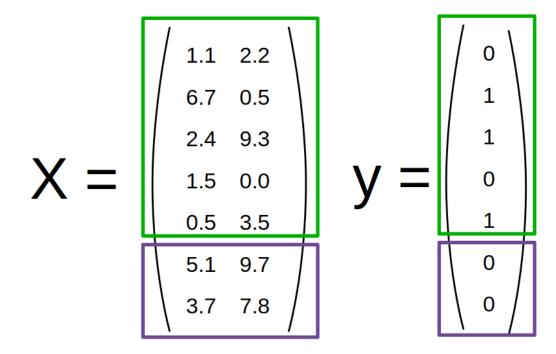
- Target y is discrete
- Example: Whether a tumor is malignant or benign based on tumour size





# Train-Test Split

#### training set



test set



### Logistic Regression

- Example: "Gender Recognition by Voice"
- Output: 0 Male and 1 Female
- Type of Problem: Binary Classification Problem
- What's the most basic thing you can do?: Weighted sum

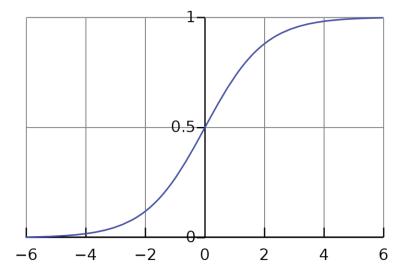
$$h_{\theta}(x) = \theta_0 x_0 + \theta_1 x_1 + \dots + \theta_j x_j \qquad = \theta^T x \qquad = \begin{bmatrix} \theta_0 & \theta_1 & \dots & \theta_j \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ \dots \\ x_j \end{bmatrix}$$

Where 
$$x_0 = 1$$



### Logistic Regression

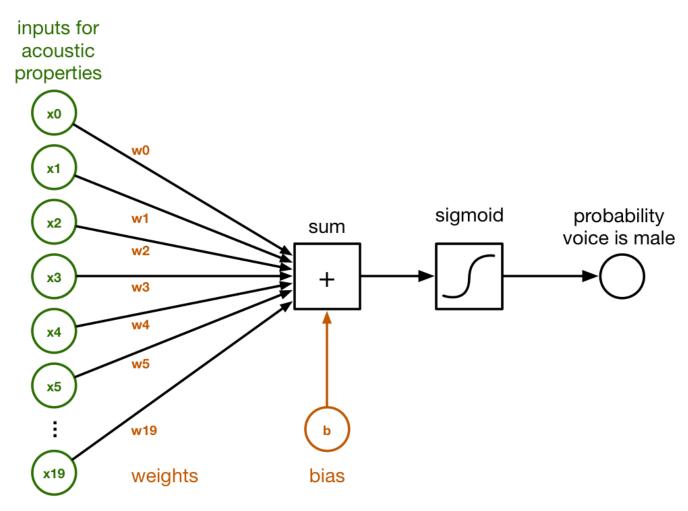
- **Next step:** "What function would help our classifier to output values between 0 and 1"
- Sigmoid Function:



• What does y axis represent?: outputs a number, we treat that value as the estimated probability that y=1. If p>0.5 we say it's 1 else we say it's a 0.

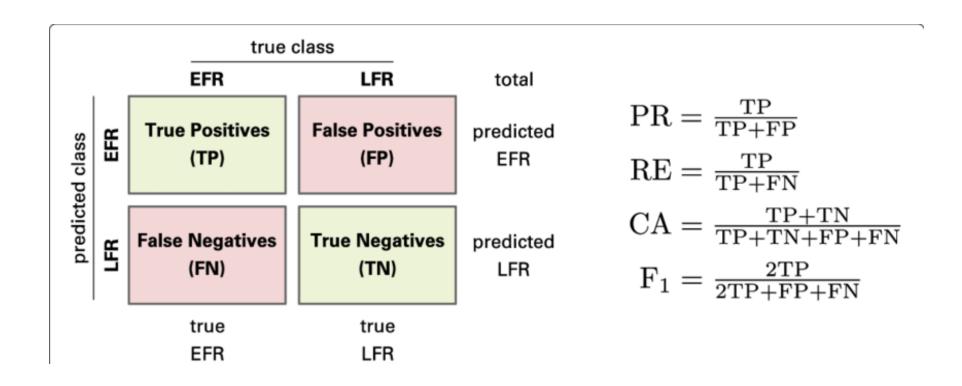


# Logistic Regression



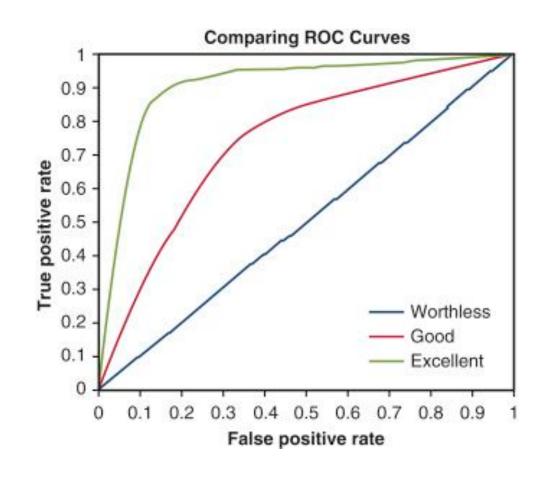


#### Model Evaluation – Confusion Matrix





### Model Evaluation – ROC and AOC





#### Loss Function

- Next Step: How to find thetas?
- **Intuition:** "This is the cost you want the learning algorithm to pay if the outcome is  $h\vartheta(x)$  and the actual outcome is y"
- Binary Cross Entropy

$$BCE = -\frac{1}{N} \sum_{i=0}^{N} y_i \cdot log(\hat{y}_i) + (1 - y_i) \cdot log(1 - \hat{y}_i)$$

- Yi is the label and ycapi is the probability obtained from the model
- This is a non-convex function i.e., there are many local optimum.



#### Independent variable

Default/Churn (Attrition Rate)

#### **Dependent variable**

**Enrollment Date** 

Price

Downpayment

Months Due

Payment Type

Use

Age

Gender



Q1: Calculate basic stats descriptive statistics (mean, median, min, max, standard deviation) for each field

• Write a function, Use Numpy, Use Pandas - describe

**Q2:**Visualize distributions of data elements using histograms for key variables and predict which variables you expect to be most correlated with default/churn.

•Matplotlib and Seaborn - distplot, pairplot

**Q3:** Calculate Linear and/or Logistic Regression Models to Predict Churn/Retention, you may choose to identify groups within the data to narrow your focus on.

- Perform Train test split of 70 30 % using train test split function in scikit learn
- Scale the data and create one hot encoded variables if required
- Use logistic regression from scikit learn
- Fit the model and predict the values using .fit and .predict



- •Display the output visually using charts of your choosing and explain your choice. (ROC Curve, Confusion Matrix, Gains Table)
  - Use confusion\_matrix and classification\_report from scikit learn
  - •Plot ROC Curve <a href="https://scikit-learn.org/stable/auto-examples/model-selection/plot-roc.html">https://scikit-learn.org/stable/auto-examples/model-selection/plot-roc.html</a>
- •In addition to the spreadsheet/code or programming output you submit, include a separate written document of 250-500 words that summarizes your process.
- •Discuss why certain variables you expected to be significant are/are not and any other unexpected insights.



- •Eliminate records with Age < 16 and Age =99 (reason age noisy due to customer not giving up age, and records incomplete where age is 99)
- •Either create an indicator variable or eliminate records where Price = Down payment (take paid in full records out of analysis for model build)
- •Only use records with date <=1998 (eliminate or set aside 1999 records since they dropped a payment type in that last year)
- •Feature possible creation (% down payment I think this would be a cool added variable but not sure if it is helpful with some modeling techniques)



# Further Reading

- Code for Logistic Regression
  - •https://chrisalbon.com/machine learning/model evaluation/plot the receiving operating characteris tic curve/
  - •https://medium.com/@kgpvijaybg/logistic-regression-on-iris-dataset-48b2ecdfb6d3
  - •https://www.kaggle.com/mnassrib/titanic-logistic-regression-with-python
  - •<u>https://medium.com/jovianml/predicting-survival-of-titanic-passengers-using-logistic-regression-model-14b9559dc4b5</u>

