

# Day 5: Mini-Project

## Summer STEM: Machine Learning

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Brooklyn, New York

June 26, 2020

# Outline

- 1 Logistic Regression
- 2 Demo: Diagnosing Breast Cancer
- 3 Multiclass Classification
- 4 Mini Project

# Classification Vs. Regression

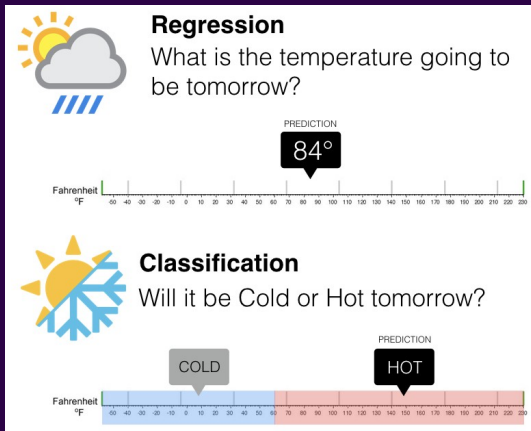
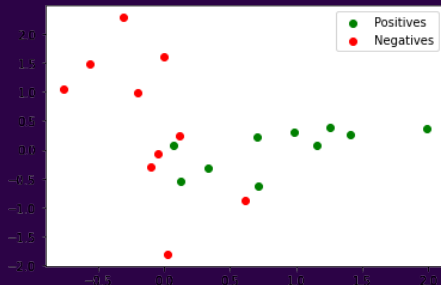


Figure: <https://www.pinterest.com/pin/672232681855858622/?lp=true>

# Classification

Given the dataset  $(x_i, y_i)$  for  $i = 1, 2, \dots, N$ , find a function  $f(x)$  (model) so that it can predict the label  $\hat{y}$  for some input  $x$ , even if it is not in the dataset, i.e.  $\hat{y} = f(x)$ .

- Positive :  $y = 1$
- Negative :  $y = 0$



## Classification via regression

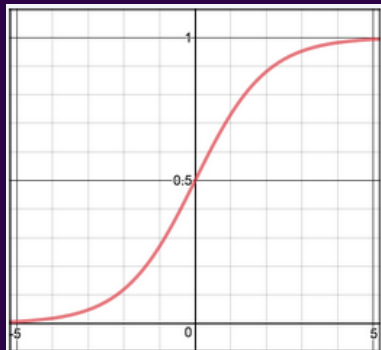
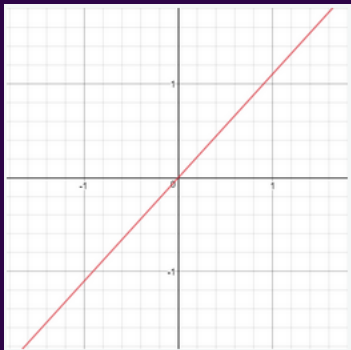
- Proposal: train a model to fit the data with linear regression!

# Classification via regression

- Proposal: train a model to fit the data with linear regression!
- What could be the problem?

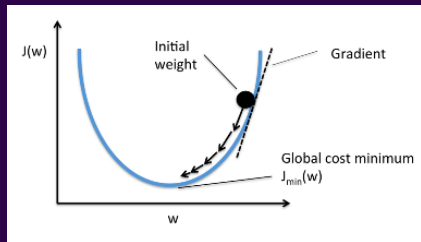
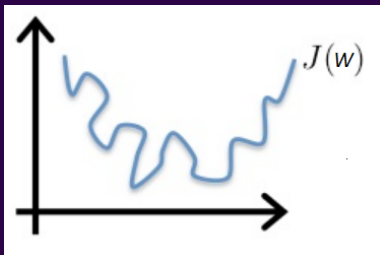
# Sigmoid Function

- Recall from linear regression  $z = w_0 + w_1x$
- By applying the sigmoid function to  $z$ , we enforce  $0 \leq \hat{y} \leq 1$ 
  - $\hat{y} = \text{sigmoid}(z) = \frac{1}{1+e^{-z}}$



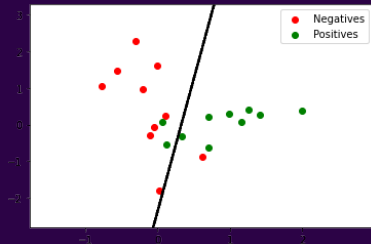
# Classification Loss Function

- Cannot use the same cost function that we used for linear regression
  - MSE of a logistic function has many local minima
- Use  $\frac{1}{N} \sum_{i=1}^N \left[ -y \log(\hat{y}) - (1 - y) \log(1 - \hat{y}) \right]$ 
  - This loss function is called binary cross entropy loss
  - This loss function has only one minimum





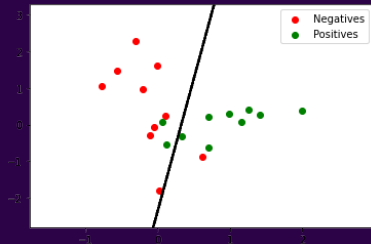
# Decision Boundary



- Evaluation metric :

$$\text{Accuracy} = \frac{\text{Number of correct prediction}}{\text{Total number of prediction}}$$

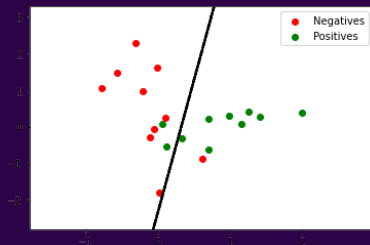
# Decision Boundary



- Evaluation metric :

$$\text{Accuracy} = \frac{\text{Number of correct prediction}}{\text{Total number of prediction}}$$

- What is the accuracy in this example ?



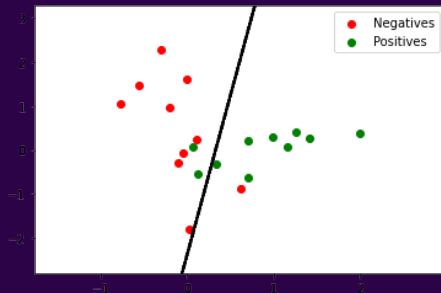
■ Evaluation metric :

$$\text{Accuracy} = \frac{\text{Number of correct prediction}}{\text{Total number of prediction}} = \frac{17}{20} = 0.85 = 85\%$$

## Types of Errors in Classification

- Correct predictions:
  - True Positive (TP) : Predict  $\hat{y} = 1$  when  $y = 1$
  - True Negative (TN) : Predict  $\hat{y} = 0$  when  $y = 0$
- Two types of errors:
  - False Positive/ False Alarm (FP):  $\hat{y} = 1$  when  $y = 0$
  - False Negative/ Missed Detection (FN):  $\hat{y} = 0$  when  $y = 1$

# Example



- How many True Positive (TP) are there ?
- How many True Negative (TN) are there ?
- How many False Positive (FP) are there ?
- How many False Negative (FN) are there ?

# Performance metrics for a classifier

- Accuracy of a classifier:
  - $(TP + TF)/(TP+FP+TN+FN)$  (percentage of correct classification)
- Why accuracy alone is not a good measure for assessing the model

# Performance metrics for a classifier

- Accuracy of a classifier:
  - $(TP + TF)/(TP+FP+TN+FN)$  (percentage of correct classification)
- Why accuracy alone is not a good measure for assessing the model
  - There might be an overwhelming proportion of one class over another (unbalanced classes)
  - Example: A rare disease occurs 1 in ten thousand people
  - A test that classifies everyone as free of the disease can achieve 99.999% accuracy when tested with people drawn randomly from the entire population

# Other metrics

## Some other metrics

- Sensitivity/Recall/TPR =  $TP / (TP + FN)$  (How many positives are detected among all positive?)
- Precision =  $TP / (TP + FP)$  (How many detected positives are actually positive?)
- Specificity/TNR =  $TN / (TN + FP)$  (How many negatives are detected among all negatives?)

Exercise: think of tasks for which sensitivity, precision, or specificity is a better metric.



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# Presentation Template

- Slide 1: Title and introduction
- Slide 2: Your model and loss function
- Slide 3 & 4: What is your choice of feature transformation, regularizer (Ridge/Lasso?) hyper-parameters, etc.
- Slide 5: Model performance on training and test set?
- Slide 6: Challenges and how you resolve them.
- Slide 7: Conclusion

# Thank You!

- Next Week: Deep Learning
- Have a fun weekend!
- Revise Revise Revise!