Nearest Neighbors Methods

Review: Types of Classifiers

• A classifier is a function that assigns to a sample, x a class label \hat{y}

$$\hat{y} = f(\mathbf{x})$$

- A probabilistic classifier obtains conditional distributions $Pr(Y|\mathbf{x})$, meaning that for a given $\mathbf{x} \in X$, they assign probabilities to all $y \in Y$
 - Hard classification

$$\hat{y} = \arg\max_{y} \Pr(Y = y | \mathbf{x})$$

Any other classifiers not belonging to a probabilistic approach?

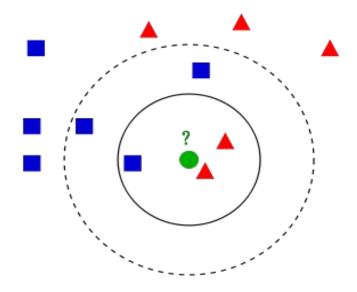
k-Nearest Neighbor

k-Nearest Neighbors(kNN)

Nonparametric method used for classification and regression

For classification

- Output class of data sample is determined by output class of its k-nearest neighbors
- Majority vote
 - \checkmark assign the output class to the most common class among k-nearest neighbors



For regression

- Output value of data sample is determined by output value of its k-nearest neighbors of the data sample
- Output value is the average value of k-nearest neighbors
 - ✓ There are several different ways to calculate average

X What is Nonparametric Method

Parametric

- Assume that data are drawn from a specific form of function up to unknown parameters
 - ✓ Linear regression, logistic regression

Nonparametric

- Assume that data are drawn from a certain unspecified function
- Unlike parametric methods, there is no single global model
- Learn to find patterns from training set and interpolate
- Heavier computational cost than parametric ones

Distance is a numerical description of how far apart objects are

- Euclidean distance, one of distance measures, is common
 - ✓ Euclidean distance of two-dimensional data points, (x_1, y_1) , (x_2, y_2)

$$d = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

✓ In general, Euclidean distance of two data points, $(x_1, x_2, ..., x_n), (y_1, y_2, ..., y_n) \in \mathbb{R}^n$

$$d = \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2 + \dots + (x_n - y_n)^2}$$

- Other distance measures
 - 1-norm distance(Manhattan distance)

$$\sum_{i}^{n} |x_i - y_i|$$

• p-norm distance (when $p=2 \rightarrow$ Euclidean distance)

$$\left(\sum_{i}^{n}(x_{i}-y_{i})^{p}\right)^{1/p}$$

Distance measure should hold the following

- $d(x,y) \ge 0$
 - ✓ Non-negativity
- $d(x,y) = 0 \Leftrightarrow x = y$
 - ✓ Identity of indiscernibles
- d(x,y) = d(y,x)
 - √ symmetry
- $d(x,z) \le d(x,y) + d(y,z)$
 - ✓ Subaddivity or triangle inequality

- What if variables are not numerical
 - Other metrics are required for categorical variables
- Metrics for categorical variables
 - Hamming distance

$$d(x,y) = \frac{\sum_{i} I(x_i \neq y_i)}{\dim(x)}$$

- ✓ $I(x_i \neq y_i)$ is 1 if and only if $x_i \neq y_i$
- \checkmark dim(x) is the dimension of x

Metrics for categorical variables

- Jaccard distance
 - ✓ Used to calculate the distance between binary vectors

	y				
		0	1		
x	0	а	b		
	1	С	d		

- \checkmark a: the total number of attributes where x and y both have a value of 0
- \checkmark b: the total number of attributes where the attribute of x is 0 and the attribute of y is 1
- \checkmark c: the total number of attributes where the attribute of x is 1 and the attribute of y is 0
- \checkmark d: the total number of attributes where x and y both have a value of 1

$$d(x,y) = \frac{b+c}{b+c+d}$$

Question

■ Find k-nearest neighbors based on given data points

1) Find k-nearest neighbors of 5^{th} objects when k=3 using Euclidean distance

2) Find k-nearest neighbors of 5^{th} objects when k=3 using Manhattan distance

index	x	у	
1	1	1	
2	2	3	
3	4	6	
4	3	1	
5	2	4	
6	4	0	
7	7	5	
8	6	2	

Feature Scaling

- Scale of variable affects on determination of nearest neighbors
- Which sample is the nearest neighbor of data sample 1?

i	x_1	x_2	x_3	x_4	у
1	9	30	100	0.5	1
2	9	25	250	0.1	0
3	9	44	220	0.7	0
4	7.5	75	170	1.2	1
	•••	•••		•••	•••



i	Distance from p_1	
1	-	
2	150.0838	
3	120.8141	
4	83.23305	

- Scale of variable x_3 dominates over other variables
- The nearest neighbor is strongly dependent on x_3

It is unfair!

Normalization

- Normalization is to adjust values of variables with different scales to common scale
 - There are several different ways for normalization
- Commonly used normalization method

$$x \to \frac{x - \mu}{\sigma}$$

- μ =mean value of the variable
- σ =standard deviation of the variable
- μ and σ are computed by sample data points

$$x \to \frac{x - x_{min}}{x_{max} - x_{min}}$$

- x_{max} is the maximum value of variable x and x_{min} is the minimum value of variable x
- Normalized value is within [0, 1]

Mahalanobis Distance

- Normalization based on normal distribution($x \to \frac{x-\mu}{\sigma}$) assumes that the sample points are distributed about the center of mass in a spherical manner
 - In real data, variables are correlated with other variables

Need to consider scale (level of spread along axis) and correlation to measure distance



Mahalonobis distance

Mahalanobis Distance

Mahalanobis distance

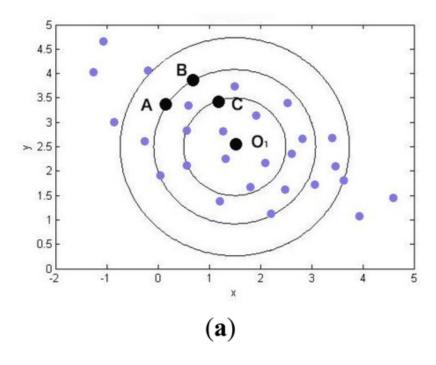
$$d(\mathbf{x}_1, \mathbf{x}_2) = \sqrt{(\mathbf{x}_1 - \mathbf{x}_2)^T S^{-1} (\mathbf{x}_1 - \mathbf{x}_2)}$$

- *S* is sample covariance matrix
- If covariance matrix is diagonal(no correlation), the resulting distance measure is as the same as the standardized distance

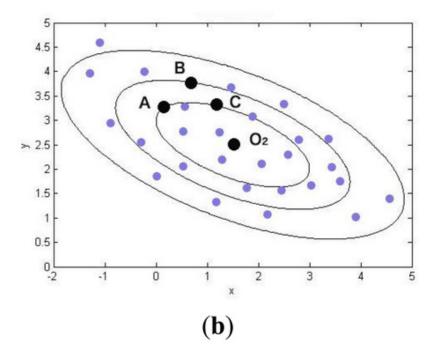
$$d(\mathbf{x}_1, \mathbf{x}_2) = \sqrt{\sum_{i=1}^{p} \frac{(x_{1i} - x_{2i})^2}{s_i^2}}$$

Mahalanobis Distance

Comparison between Euclidean distance and Mahalanobis distance



Euclidean distance



Mahalanobis distance

Procedure of k-NN

Decide the number of nearest neighbors k and distance measure

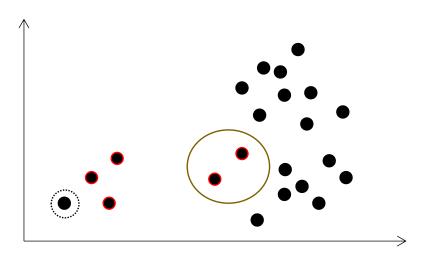
For all data point in test set, find k nearest neighbors

Obtain output value based on output values of neighbors

Fixed-radius Nearest Neighbors

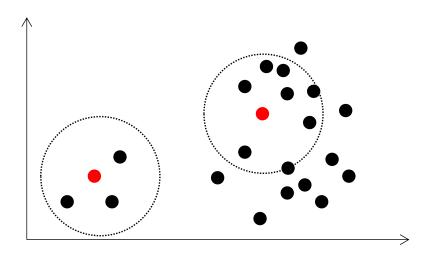
Problem of Fixed-Number of Nearest Neighbors

- When distribution of data set is not homogenous, samples not similar to data point \boldsymbol{x} can be obtained in the nearest neighbors
 - k = 5



Fixed-Radius Near Neighbors

- Fixed-radius near neighbors are neighbors within fixed range from data point \boldsymbol{x}
 - Because of that, the number of neighbors may be different depending on the location



Fixed-Radius Near Neighbors Methods

- The only difference of fixed-radius NN from kNN is the method to find the nearest neighbors
 - Remained steps of classification and regression are the same

Decide radius of range from data point and distance measure

For all data point in test set, find fixed-radius near neighbors

Obtain output value based on output values of neighbors

Thank you! Thank you!