

MSBA 7027 Machine Learning

K-Nearest Neighbors

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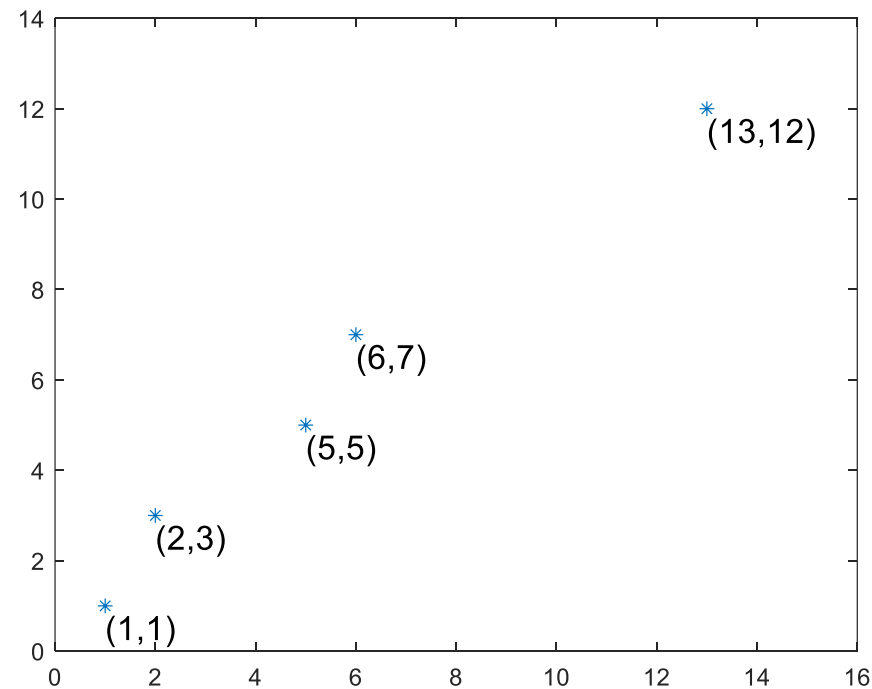
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K-Nearest Neighbors (KNN)

- Non-parametric method
 - Can be used for both regression & classification problems
- Given K and a prediction point $x^{(n+1)}$,
 - KNN identifies the closest K training observations
 - Then estimate $f(x^{(n+1)})$ using their averages

K-Nearest Neighbors (KNN)

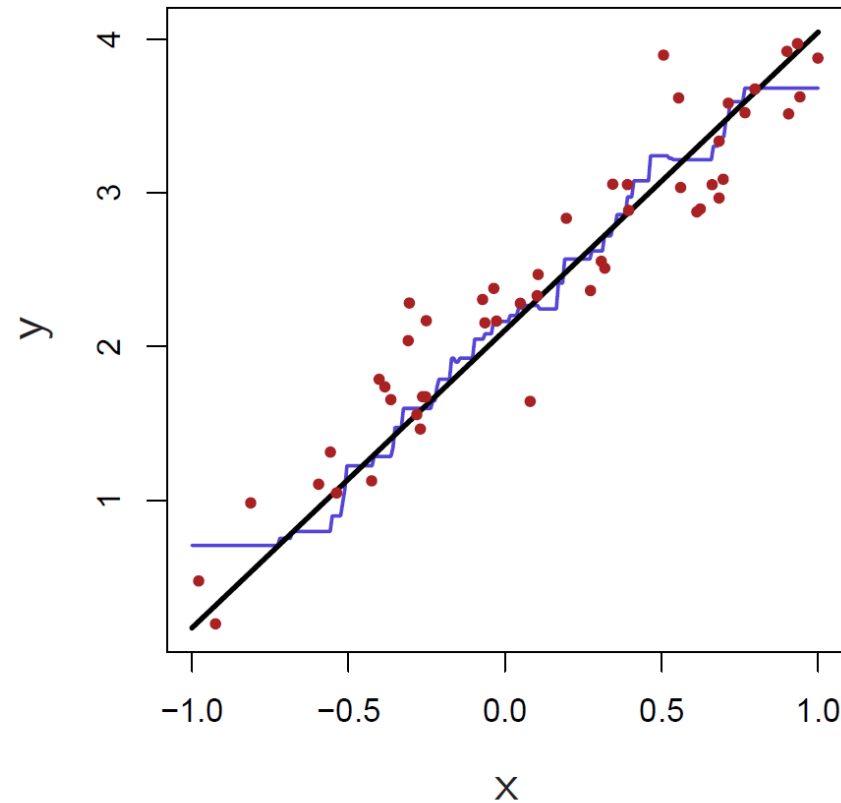
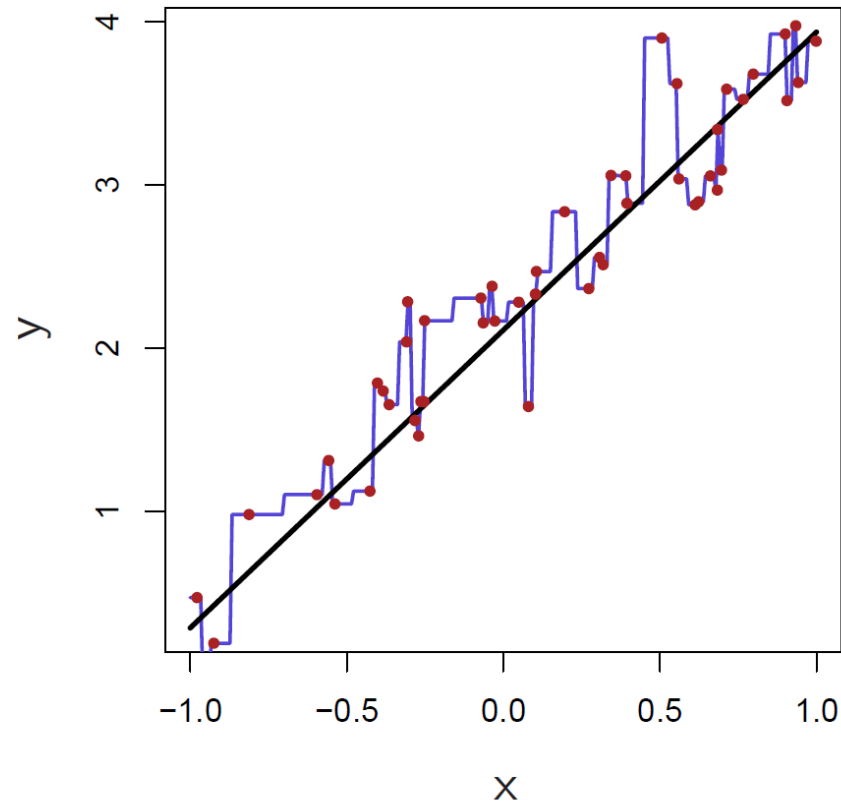
- One-dimensional Example
 - $(x^{(i)}, y^{(i)}) = (1, 1), (2, 3), (5, 5), (6, 7), (13, 12)$
 - $K = 3$



K-Nearest Neighbors (KNN)

- In practice, choose K by cross-validation (minimize SSE in test set)
 - Why not minimize SSE in training set (like in OLS)
- Value of K : Bias-variance tradeoff
- Small value of K vs large value of K

K-Nearest Neighbors (KNN)



Which one has a smaller K?

K-Nearest Neighbors (KNN)

- Features multi-dim: Euclidean distance
- Classification problem: Majority vote

K-Nearest Neighbors (KNN)

- Sample final exam question
 - $x^{(i)} = (x_1^{(i)}, x_2^{(i)})$
 - $(x^{(i)}, y^{(i)}) = (0, 3, 5), (4, 0, 1), (8, 3, 7)$
 - $K = 2$
- Given $x^{(4)} = (2, 2)$, what is \hat{y} ?
- Derive \hat{y} over the whole plane of (x_1, x_2) .

KNN – implementation in R

Getting started

Load relevant packages

```
# Example of installing a package  
install.packages('FNN')
```

```
library(FNN) # For func knn  
library(gmodels) # For func CrossTable
```

Getting started

Load Data

```
# Loading iris dataset  
iris.rawData <- iris
```

```
# Viewing iris dataset structure and attributes  
summary(iris.rawData)
```

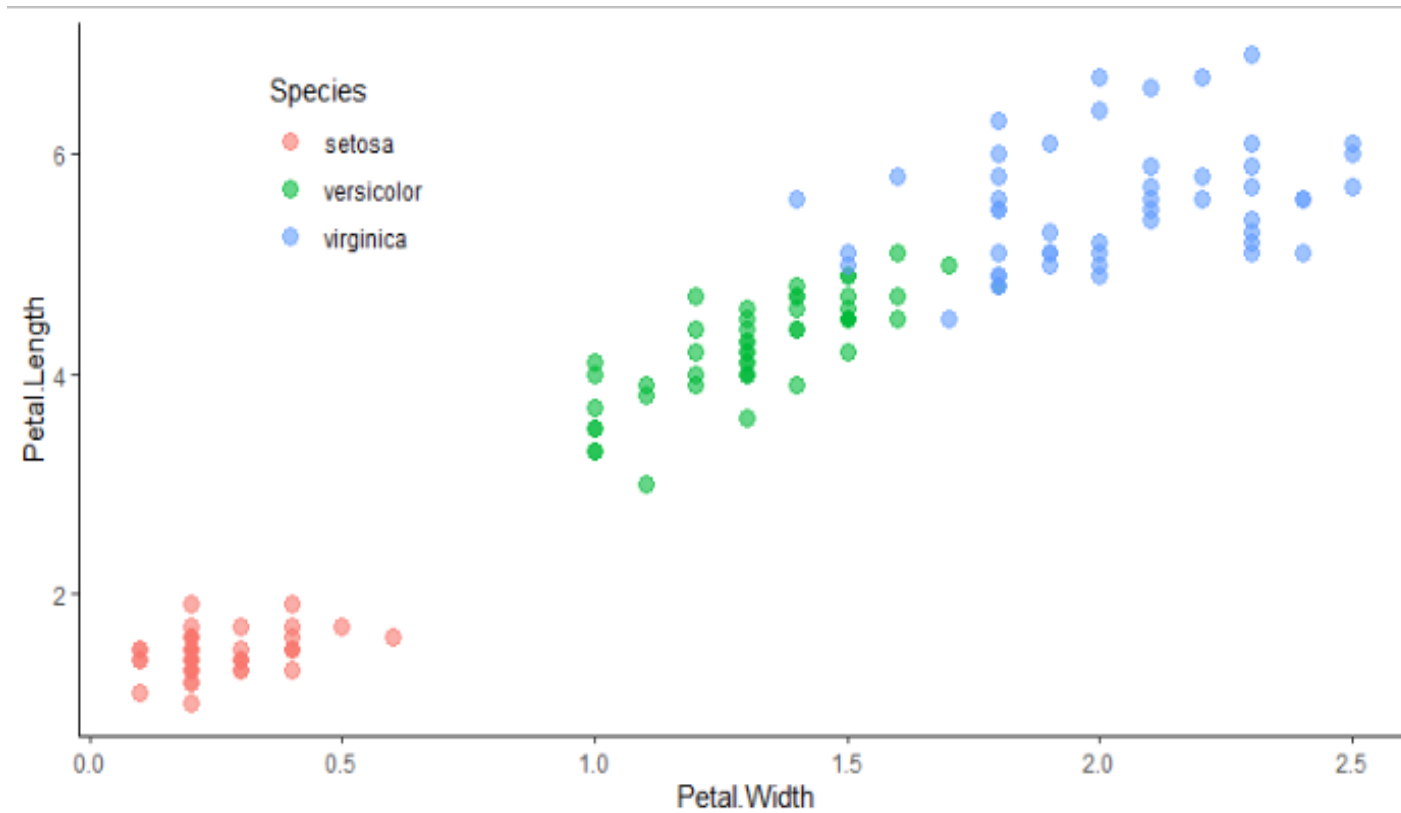
```
> summary(iris.data)
```

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
Min. :4.300	Min. :2.000	Min. :1.000	Min. :0.100	setosa :50
1st Qu.:5.100	1st Qu.:2.800	1st Qu.:1.600	1st Qu.:0.300	versicolor:50
Median :5.800	Median :3.000	Median :4.350	Median :1.300	virginica :50
Mean :5.843	Mean :3.057	Mean :3.758	Mean :1.199	
3rd Qu.:6.400	3rd Qu.:3.300	3rd Qu.:5.100	3rd Qu.:1.800	
Max. :7.900	Max. :4.400	Max. :6.900	Max. :2.500	

3 Classes of Iris Species: Setosa, versicolor, virginica

Data visualisation

Features: Petal.Width, Petal.Length



Data Standardization

standardize data

```
standardize <- function(x) {  
  return ( x - mean(x) )/( sd(x) )  
}
```

Only standardize the first 4 columns (5th column is label)

```
iris.standardizeData = iris.rawData  
for(i in seq(1,4)){  
  iris.standardizeData[,i] = standardize(iris.rawData[,i])  
}
```

Split into train & test set

```
set.seed(123)  
split <- rsample::initial_split(iris.standardizeData, prop = 0.7, strata = "Species")  
iris.train <- rsample::training(split)  
iris.test <- rsample::testing(split)  
iris.trainFeatMat = iris.train[,1:4]  
iris.trainLabel <- iris.train[,5]  
iris.testFeatMat <- iris.test[,1:4]  
iris.testLabel <- iris.test[,5]
```

Perform KNN

knn function with the following parameters:

- Train feature matrix
- Test feature matrix
- Train labels
- A value for K

Returns predicted test labels

Building our knn classifier

```
predictTestLabel <- knn(train = iris.trainFeatMat, test = iris.testFeatMat, cl = iris.trainLabel, k = 3)
```

Model Evaluation

Contingency Table

CrossTable(x = iris.testLabel, y = predictTestLabel, prop.chisq = FALSE)

- **Overall, KNN performs very well**
 - Able to predict correctly almost all instances
 - Only misclassifies two instances

Total Observations in Table: 45

iris.testLabel	predictTestLabel			Row Total
	setosa	versicolor	virginica	
setosa	15	0	0	15
	1.000	0.000	0.000	0.333
	1.000	0.000	0.000	
	0.333	0.000	0.000	
versicolor	0	15	0	15
	0.000	1.000	0.000	0.333
	0.000	0.882	0.000	
	0.000	0.333	0.000	
virginica	0	2	13	15
	0.000	0.133	0.867	0.333
	0.000	0.118	1.000	
	0.000	0.044	0.289	
Column Total	15	17	13	45
	0.333	0.378	0.289	

Perform KNN

Note: For KNN-regression, syntax is the same except changing the function name to: `knn.reg`

End