## NC STATE UNIVERSITY

# Nonlinear Methods: k Nearest Neighbors

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## **Supervised Learning**

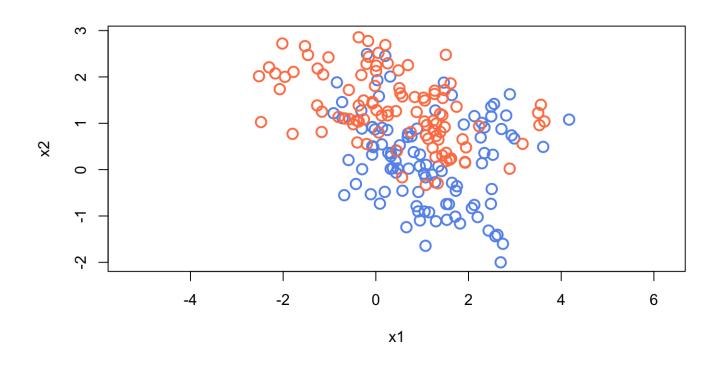
Prediction our goal:

Generally broken into two categories

- Regression
  - Quantitative response
  - Prediction of response
- Classification
  - Categorical response
  - Predict class membership or probability of membership
  - Saw logistic regression and will see classification trees, now we'll look at K Nearest Neighbors (KNN)

## Example

Suppose you have two predictors and a categorical response (red or blue)

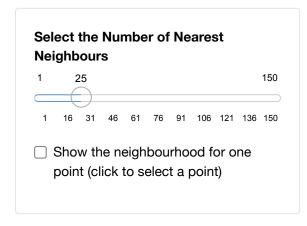


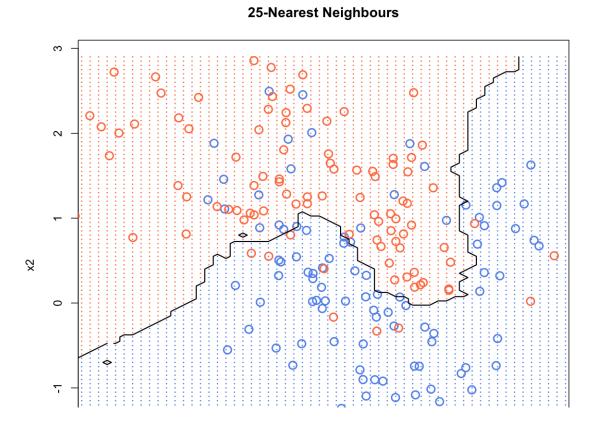
#### **kNN**

Want to predict class membership (red or blue) based on (x1, x2) combination k Nearest Neighbor idea:

- Use "closest" k observations from training set to predict class (should usually standardize predictors - center/scale)
- Often use Euclidean distance between predictors to determine closest
- P(red|x1, x2) = proportion of k closest values that are red
- P(blue|x1,x2) = proportion of k closest values that are blue
- · Classify (predict) to class with highest probability

#### k-Nearest Neighbours Classification

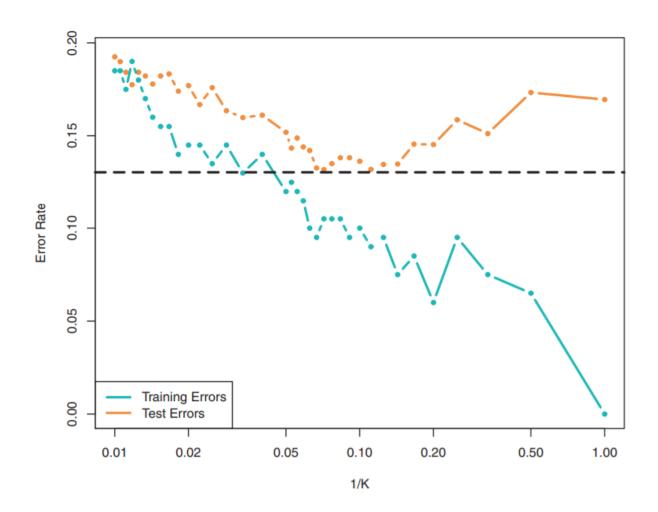




#### **kNN**

- Small k implies flexible (possibly overfit, higher variance)
  - Training error will be small, may not extend to testing error
- Large k implies more rigid (possibly underfit, lower variance)
- · Can do training and test or CV to determine k

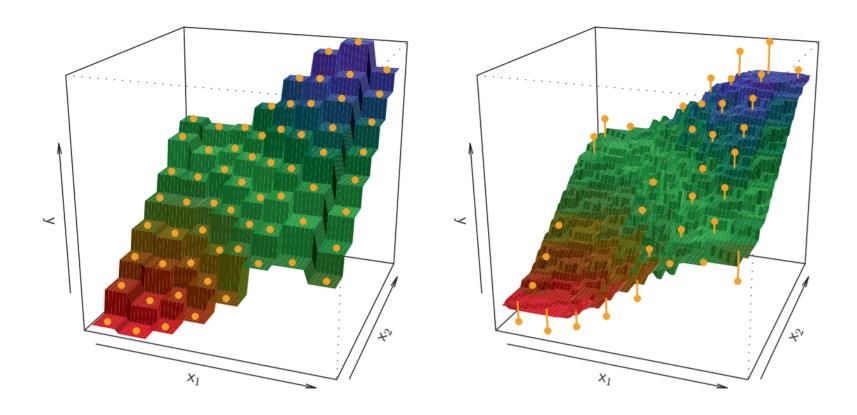
Judge error using misclassification rate.



Courtesy: Introduction to Statistical Learning

## **kNN** for Regression

- · Same idea, use average of responses of "closest" k observations in training set as prediction
- · Closest again often Euclidean distance
- Very flexible



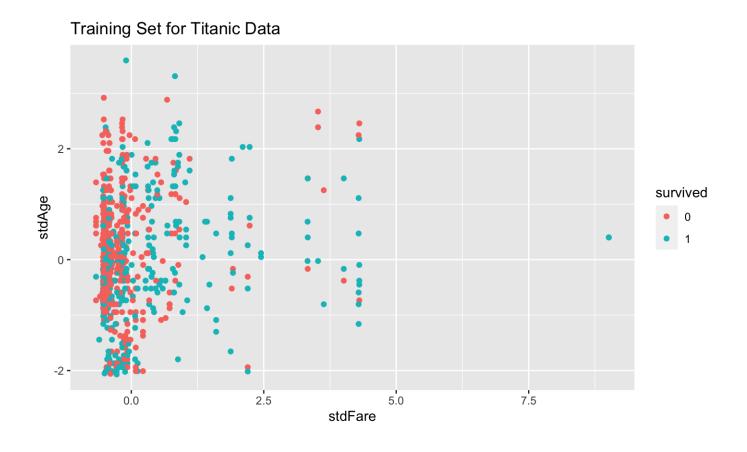
Courtesy: Introduction to Statistical Learning k = 1 on the left, k = 9 on the right

#### Fitting kNN in R with class::knn() Function

- Titanic Data set: Predict survival status (removed NAs) as a function of traveler age and traveler fare
- Should center and scale (divide by SD) data
  - Use training mean and training SD to do standardization for both training and test set

## Fitting kNN in R with class::knn() Function

 Titanic Data set: Predict survival status (removed NAs) as a function of traveler age and traveler fare



#### kNN in R

```
knnFit <- knn(train = select(titanicDataTrain, stdFare, stdAge),</pre>
                          test = select(titanicDataTest, stdFare, stdAge),
                           cl = titanicDataTrain$survived,
                          k = 3) #could use CV to determine k
fitInfo <- tbl df(data.frame(knnFit, select(titanicDataTest, survived, stdFare, stdAge)))
## Warning: `tbl df()` was deprecated in dplyr 1.0.0.
## Please use `tibble::as tibble()` instead.
fitInfo
## # A tibble: 209 x 4
## knnFit survived stdFare stdAge
## <fct> <fct> <dbl> <dbl>
## 1 1
                     2.20 0.0464
           0
## 2 0
              0.269 2.96
## 3 1
              0.776 0.188
## 4 1
                     3.63 0.898
           1
## 5 1
                      1.05 - 0.308
## # ... with 204 more rows
```

#### kNN in R

```
tbl1 <- table(fitInfo$knnFit,fitInfo$survived)
tbl1

##
## 0 1
## 0 84 39
## 1 44 42

#misclass rate in test set
misClass <- 1 - sum(diag(tbl1))/sum(tbl1)
misClass
## [1] 0.3971292</pre>
```

## kNN vs Logistic Regression

## kNN vs Logistic Regression

kNN does very poorly here - perhaps choose k with CV!

```
## Logistic Regression Misclassification Rate
## 0.3301435
## kNN Misclassification Rate
## 0.3971292
## Predicting Death for All
## 0.3875598
```

#### With k = 10

```
##
##
        0 1
##
     0 98 41
     1 30 40
##
## [1] 0.3397129
## Logistic Regression Misclassification Rate
##
                                    0.3301435
##
                   kNN Misclassification Rate
##
                                    0.3397129
##
                     Predicting Death for All
##
                                    0.3875598
```

#### Recap

- k nearest neighbors uses close observations from the training set for prediction
- Very flexible to not flexible
- · Can be used for both regression and classification