Titanc

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Contents

Data Collection:

There is two type of data set

Exploring Data:

The trained dataset contains 891 observations and 12 features (Variable), and the tested dataset contains 418 observations.

```
> str(train.df)
'data.frame': 891 obs. of 12 variables:
$ PassengerId: int 1 2 3 4 5 6 7 8 9 10 ...
```

```
$ Survived
            : int 0 1 1 1 0 0 0 0 1 1 ...
$ Pclass
            : int 3 1 3 1 3 3 1 3 3 2 ...
            : chr "Braund, Mr. Owen Harris" "Cumings, Mrs. John Bradley (Florer
$ Name
$ Sex
            : chr "male" "female" "female" "female" ...
            : num 22 38 26 35 35 NA 54 2 27 14 ...
$ Age
$ SibSp
            : int 1 1 0 1 0 0 0 3 0 1 ...
            : int 000000120 ...
$ Parch
            : chr "A/5 21171" "PC 17599" "STON/O2. 3101282" "113803" ...
$ Ticket
$ Fare
            : num 7.25 71.28 7.92 53.1 8.05 ...
$ Cabin
            : chr NA "C85" NA "C123" ...
                   "S" "C" "S" "S" ...
$ Embarked
            : chr
```

Pclass need to be converted in factor

- > train.df\$Pclass = as.factor(train.df\$Pclass)
- > test.df\$Pclass = as.factor(test.df\$Pclass)

Preparion Data:

Missing Data Analysis

We have to analyse is there any missing data.

Train Data:

> sapply(train.df, function(x) sum(is.na(x)))

PassengerId	Survived	Pclass	Name	Sex	Age
0	0	0	0	0	177
SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	0	0	0	687	2

Test Data:

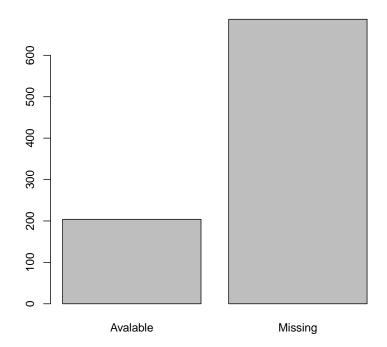
> sapply(test.df, function(x) sum(is.na(x)))

PassengerId	Pclass	Name	Sex	Age	SibSp
0	0	0	0	86	0
Parch	Ticket	Fare	Cabin	Embarked	Survived
0	0	1	327	0	0

Age and cabin need to handle. Lets check what fraction of data is missing in both.

Cabin in train data set

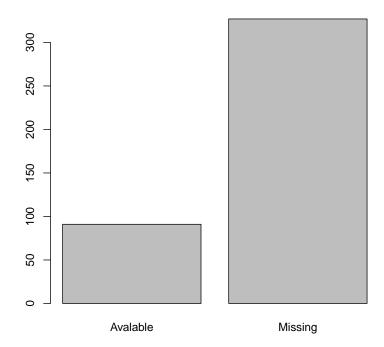
Train Cabin Missing vs Avalable Data



Cabin in test data set

```
> test.df['CabinMissing'] = sapply(test.df$Cabin,
+ function(x)
+ ifelse(is.na(x),
```

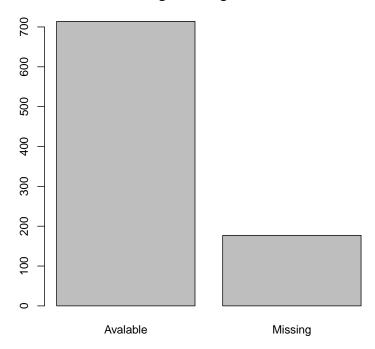
Train Cabin Missing vs Avalable Data



Age in train dataset

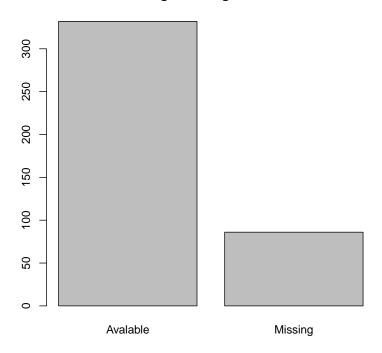
```
> train.df['AgeMissing'] = sapply(train.df$Age,
+ function(x)
+ ifelse(is.na(x),
+ "Missing", "Avalable")
+ )
> barplot(table(factor(train.df$AgeMissing)),
+ main = "Train Age Missing vs Avalable")
```

Train Age Missing vs Avalable



Age in test dataset

Train Age Missing vs Avalable



Since cabin have many missing data we can remove this feather from out data set.

> train.df = subset(train.df, select = c(2,3,5,6,7,8,10,12))

```
> test.df = subset(test.df, select = c(2,4,5,6,7,9,11,12))
> names(train.df)
[1] "Survived" "Pclass"
                           "Sex"
                                      "Age"
                                                  "SibSp"
                                                              "Parch"
                                                                         "Fare"
[8] "Embarked"
> names(test.df)
[1] "Pclass"
                           "Age"
                                       "SibSp"
                                                                         "Embarked"
               "Sex"
                                                  "Parch"
                                                              "Fare"
```

To handle missing value of Age we can apply one of these method.

• Throw out any data with missing values

[8] "Survived"

- Assign the average value
- Use a regression or another simple model to predict the values of missing variables

For checking weather we can use regression or not lets check correlation in predicter.

No correlation found in Age and other predictor. hence regression is not used for missing age. we will fill age by avarage value.

```
> agetrain = train.df$Age
> avgTrainAge = mean(agetrain, na.rm = T)
> train.df$Age[is.na(train.df$Age)] = avgTrainAge
> ageTest = test.df$Age
> avgTestAge = mean(ageTest, na.rm = T)
> test.df$Age[is.na(test.df$Age)] = avgTestAge
>
```

We can remove row with missing

```
> train.df = train.df[!is.na(train.df$Embarked),]
> test.df = test.df[!is.na(test.df$Fare),]
```

Now again checking missing value count

Train data

> sapply(train.df, function(x) sum(is.na(x)))

Survived	Pclass	Sex	Age	SibSp	Parch	Fare Emba	arked
0	0	0	0	0	0	0	0

Test data

> sapply(test.df, function(x) sum(is.na(x)))

Pclass	Sex	Age	SibSp	Parch	Fare	${\tt Embarked}$	Survived
0	0	0	0	0	0	0	0

Model on Train data

Logistic Regression

```
> glm1 = glm(train.df$Survived ~., family = binomial(link = 'logit'), data = tra
> summary(glm1)
```

Call:

Deviance Residuals:

```
Min 1Q Median 3Q Max -2.6235 -0.6098 -0.4237 0.6112 2.4512
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)
(Intercept) 4.102784
                      0.476303 8.614 < 2e-16 ***
Pclass2
          -0.924047
                      0.297882 -3.102 0.00192 **
Pclass3
           -2.149626
                      0.297749 -7.220 5.21e-13 ***
Sexmale
          -2.709611
                      0.201336 -13.458 < 2e-16 ***
Age
          -0.039320
                      0.007888 -4.984 6.21e-07 ***
SibSp
          -0.322143
                      0.109545 -2.941 0.00327 **
                      0.119028 -0.799 0.42450
Parch
           -0.095061
                      0.002462 0.918 0.35842
           0.002261
Fare
```

```
-0.029839
                       0.381534 -0.078 0.93766
EmbarkedQ
EmbarkedS
           -0.445754
                       0.239730 -1.859 0.06297 .
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1182.82 on 888 degrees of freedom
Residual deviance: 783.74 on 879 degrees of freedom
AIC: 803.74
Number of Fisher Scoring iterations: 5
Evauvate model
> testPredict = predict(glm1, newdata = subset(test.df, select = c(1:7)),
                           type = 'response')
> testPredict = ifelse(testPredict >0.5, 1, 0)
> cm = table(test.df[, 'Survived'], testPredict > 0.5)
> cm
   FALSE TRUE
     252
           13
      11 141
  Acuracy
> misClassifiError = mean(testPredict != test.df$Survived)
> #print(paste('Accuracy', 1 - testPredict))
> 1- misClassifiError
[1] 0.942446
```