

# Titanic (Train)

## Session 6 Assignment 1

1. Import the Titanic Dataset from the link [Titanic Data Set](#).

Perform the following:

- Preprocess the passenger names to come up with a list of titles that represent families and represent using appropriate visualization graph.
- Represent the proportion of people survived from the family size using a graph.
- Impute the missing values in Age variable using Mice Library, create two different graphs showing Age distribution before and after imputation.

a. Preprocess the passenger names to come up with a list of titles that represent families and represent using appropriate visualization graph.

### Preprocessing Data

```
library(reshape)
library(caret)
d <- train
d.nrow<-seq(1, nrow(d)) # save the number of rows in the train dataset
d.miss <- melt(apply(d[, -2], 2, function(x) sum(is.na(x) | x=="")))
cbind(row.names(d.miss)[d.miss$value>0], d.miss[d.miss$value>0,])
```

```
[,1]      [,2]
[1,] "Age"   "177"
[2,] "Cabin" "687"
[3,] "Embarked" "2"
```

```
#Variable "Cabin"
#"Cabin" has missed about 80% values. We will not use this variable.
```

```
#Variable "Embarked"
#Update missing Embarked value with the most common value:
```

```
#table(d$Embarked)
#Variable "Price"
#Some Fare values contains sum for tickets were purchased in groups. Introduce a new variable "Price" that will be Fare per person.

d$Fare[which(is.na(d$Fare))] <- 0 # Update missing Fare value with 0.
# calculate Ticket Price (Fare per person)
ticket.count <- aggregate(d$Ticket, by=list(d$Ticket), function(x) sum( !is.na(x) ))
d$Price<-apply(d, 1, function(x) as.numeric(x["Fare"]) / ticket.count[which(ticket.count[, 1] == x["Ticket"]), 2])
```

Capt	Col	Don	Dr	Jonkheer	Lady
1	2	1	7	1	1
Major	Master	Miss	Mlle	Mme	Mr
2	40	182	2	1	517
Mrs	Ms	Rev	Sir	the Countess	
125	1	6	1	1	

#Price related to passenger class. Missig price values (price=0) we can update with median price per passenger class:

```
pclass.price<-aggregate(d$Price, by = list(d$Pclass), FUN = function(x) median(x, na.rm = T))
d[which(d$Price==0), "Price"] <- apply(d[which(d$Price==0), ], 1, function(x) pclass.price[pclass.price[, 1]==x["Pclass"], 2])
#Variable "Title"
#Extract title of each persons name to a new variable "Title"
d$Title<-regmatches(as.character(d$Name),regexpr("\\\\,[A-z ]{1,20}\\\\\\.", as.character(d$Name)))
d$Title<-unlist(lapply(d$Title,FUN=function(x) substr(x, 3, nchar(x)-1)))
table(d$Title)
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

#Merge 17 different title groups to the most common 4 groups.

```
d$Title[which(d$Title %in% c("Mme", "Mlle"))] <- "Miss"
d$Title[which(d$Title %in% c("Lady", "Ms", "the Countess", "Dona"))] <- "Mrs"
d$Title[which(d$Title=="Dr" & d$Sex=="female")] <- "Mrs"
d$Title[which(d$Title=="Dr" & d$Sex=="male")] <- "Mr"
d$Title[which(d$Title %in% c("Capt", "Col", "Don", "Jonkheer", "Major", "Rev", "Sir"))] <- "Mr"
d$Title<-as.factor(d$Title) #convert to factor variable
```

```

#Variable "Age"
#Update unknown age with median age for each group of title:

title.age<-aggregate(d$Age,by = list(d$Title), FUN = function(x) median(x, na.rm = T))
d[is.na(d$Age), "Age"] <- apply(d[is.na(d$Age), ], 1, function(x) title.age[title.age[, 1]==x["Title"], 2])
#Split train and test data
#We merged train and test data at the begining of preprocess. Now we will split it back to "t" and "d" Data frame variables.
#Data frame "t" has no "Survival" values and will be used to predict "Survival" and submit on Kaggle.
#Data frame "d" that contains train data we also split to test prediction models.

t <- d[-d.nrow, ] # test data. It has no "Survival" values.
d <- d[d.nrow, ] #Train data
set.seed(1234)
inTrain<-createDataPartition(d$Survived, p = 0.8)[[1]]
#Fitting a linear model that includes all variables.
fit.8 <- glm(Survived ~ Pclass+Sex+Age+SibSp+Parch+Embarked+Title+Price+Ticket, data=d[inTrain,], family=binomial("logit"))
summary(fit.8)
#Fitting a linear model that includes 5 statistically significant variable and "Ticket" converted to a factor variable.
fit.6.grp <- glm(Survived ~ Pclass+Age+SibSp+Parch+Title+I(Ticket>2), data=d[inTrain,], family=binomial)
summary(fit.6.grp)

```

```

Call:
glm(formula = Survived ~ Pclass + Sex + Age + SibSp + Parch + Embarked + Title + Price + Ticket, family = binomial("logit"), data = d[inTrain, ])

```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-8.49	0.00	0.00	0.00	8.49

Coefficients: (4 not defined because of singularities)

	Estimate	Std. Error	z value	Pr(> z )	
(Intercept)	2.746e+16	5.354e+08	5.130e+07	<2e-16	***
Pclass	-9.834e+15	2.233e+08	-4.404e+07	<2e-16	***
Sexmale	6.646e+14	2.627e+07	2.531e+07	<2e-16	***
Age	-1.852e+12	6.684e+05	-2.771e+06	<2e-16	***
SibSp	-1.666e+14	1.213e+07	-1.373e+07	<2e-16	***
Parch	4.025e+12	1.418e+07	2.839e+05	<2e-16	***
EmbarkedQ	8.159e+15	2.468e+08	3.306e+07	<2e-16	***
EmbarkedS	-4.140e+14	6.218e+07	-6.659e+06	<2e-16	***
TitleMiss	4.578e+14	2.026e+07	2.260e+07	<2e-16	***
TitleMr	-3.269e+15	2.549e+07	-1.282e+08	<2e-16	***
TitleMrs	NA	NA	NA	NA	

Price	-4.498e+14	1.031e+07	-4.363e+07	<2e-16	***
Ticket110413	-2.885e+15	7.539e+07	-3.826e+07	<2e-16	***
Ticket110465	-5.098e+15	7.790e+07	-6.544e+07	<2e-16	***
Ticket111240	5.689e+14	7.911e+07	7.190e+06	<2e-16	***
Ticket111320	2.792e+15	1.070e+08	2.610e+07	<2e-16	***
Ticket111361	-4.506e+15	1.001e+08	-4.503e+07	<2e-16	***
Ticket111369	3.019e+15	9.892e+07	3.052e+07	<2e-16	***
Ticket111427	1.889e+15	8.795e+07	2.148e+07	<2e-16	***
Ticket112050	1.315e+15	8.708e+07	1.510e+07	<2e-16	***
Ticket112058	1.299e+15	8.823e+07	1.472e+07	<2e-16	***
Ticket112059	1.317e+15	8.698e+07	1.514e+07	<2e-16	***
Ticket112379	2.841e+15	1.332e+08	2.133e+07	<2e-16	***
Ticket113028	-2.615e+15	8.795e+07	-2.973e+07	<2e-16	***
Ticket113050	-2.587e+15	8.853e+07	-2.922e+07	<2e-16	***
Ticket113051	-2.489e+15	1.034e+08	-2.406e+07	<2e-16	***
Ticket113055	1.928e+15	8.908e+07	2.164e+07	<2e-16	***
Ticket113059	6.625e+15	1.840e+08	3.600e+07	<2e-16	***
Ticket113501	-1.065e+15	7.642e+07	-1.393e+07	<2e-16	***
Ticket113503	8.015e+16	1.855e+09	4.321e+07	<2e-16	***
Ticket113505	-2.825e+14	8.403e+07	-3.362e+06	<2e-16	***
Ticket113509	1.297e+16	3.252e+08	3.987e+07	<2e-16	***
Ticket113510	1.186e+15	8.694e+07	1.364e+07	<2e-16	***
Ticket113514	-2.555e+15	9.054e+07	-2.822e+07	<2e-16	***
Ticket113760	1.355e+15	5.812e+07	2.331e+07	<2e-16	***
Ticket113776	-7.225e+14	6.287e+07	-1.149e+07	<2e-16	***
Ticket113781	1.084e+15	8.222e+07	1.319e+07	<2e-16	***
Ticket113783	-1.000e+15	9.235e+07	-1.083e+07	<2e-16	***
Ticket113784	1.439e+15	8.785e+07	1.638e+07	<2e-16	***
Ticket113786	3.706e+15	7.562e+07	4.902e+07	<2e-16	***
Ticket113788	5.911e+15	8.989e+07	6.576e+07	<2e-16	***
Ticket113789	-3.982e+15	7.933e+07	-5.019e+07	<2e-16	***
Ticket113792	-2.566e+15	8.967e+07	-2.862e+07	<2e-16	***
Ticket113794	1.896e+15	8.799e+07	2.155e+07	<2e-16	***
Ticket113798	-9.323e+15	1.729e+08	-5.393e+07	<2e-16	***
Ticket113800	-2.560e+15	9.023e+07	-2.837e+07	<2e-16	***
Ticket113803	-2.897e+15	7.539e+07	-3.842e+07	<2e-16	***
Ticket113806	2.050e+15	8.968e+07	2.286e+07	<2e-16	***
Ticket113807	-2.555e+15	9.054e+07	-2.822e+07	<2e-16	***
Ticket11668	-3.391e+15	8.323e+07	-4.074e+07	<2e-16	***
Ticket11751	6.490e+14	8.161e+07	7.953e+06	<2e-16	***
Ticket11753	1.377e+16	2.300e+08	5.989e+07	<2e-16	***
Ticket11755	4.941e+15	1.296e+08	3.811e+07	<2e-16	***
Ticket11765	1.463e+16	2.683e+08	5.451e+07	<2e-16	***
Ticket11767	5.675e+15	1.420e+08	3.998e+07	<2e-16	***
Ticket11769	1.087e+16	2.167e+08	5.018e+07	<2e-16	***
Ticket11771	-1.560e+15	9.959e+07	-1.566e+07	<2e-16	***
Ticket11774	2.892e+15	9.902e+07	2.920e+07	<2e-16	***
Ticket11813	2.079e+16	4.747e+08	4.380e+07	<2e-16	***
Ticket11967	1.017e+16	1.773e+08	5.738e+07	<2e-16	***
Ticket12233	1.132e+15	9.822e+07	1.152e+07	<2e-16	***
Ticket12460	2.417e+13	9.509e+07	2.542e+05	<2e-16	***
Ticket12749	7.913e+15	1.768e+08	4.475e+07	<2e-16	***
Ticket13049	3.088e+15	1.361e+08	2.269e+07	<2e-16	***
Ticket13213	5.548e+15	1.075e+08	5.163e+07	<2e-16	***
Ticket13214	3.255e+15	9.825e+07	3.313e+07	<2e-16	***
Ticket13502	-4.137e+15	7.559e+07	-5.473e+07	<2e-16	***
Ticket13507	-1.782e+15	8.350e+07	-2.134e+07	<2e-16	***

Ticket13509	-2.550e+15	9.104e+07	-2.801e+07	<2e-16	***
Ticket13567	2.537e+16	4.941e+08	5.136e+07	<2e-16	***
Ticket13568	8.713e+15	2.047e+08	4.256e+07	<2e-16	***
Ticket14312	-3.054e+15	9.681e+07	-3.154e+07	<2e-16	***
Ticket14313	1.449e+15	9.681e+07	1.497e+07	<2e-16	***
Ticket14973	8.717e+15	2.414e+08	3.611e+07	<2e-16	***
Ticket1601	1.003e+16	2.327e+08	4.310e+07	<2e-16	***
Ticket16966	1.480e+16	3.777e+08	3.918e+07	<2e-16	***
Ticket16988	3.441e+15	7.635e+07	4.506e+07	<2e-16	***
Ticket17421	-2.769e+15	8.907e+07	-3.109e+07	<2e-16	***
Ticket17453	6.207e+15	1.604e+08	3.870e+07	<2e-16	***
Ticket17464	1.085e+16	2.265e+08	4.789e+07	<2e-16	***
Ticket17465	-9.591e+14	9.166e+07	-1.046e+07	<2e-16	***
Ticket17466	-9.610e+14	9.156e+07	-1.050e+07	<2e-16	***
Ticket17474	-6.323e+14	6.465e+07	-9.780e+06	<2e-16	***
Ticket17764	-1.116e+15	9.830e+07	-1.135e+07	<2e-16	***
Ticket19877	1.731e+15	1.016e+08	1.703e+07	<2e-16	***
Ticket19928	-4.154e+15	1.757e+08	-2.364e+07	<2e-16	***
Ticket19943	6.813e+15	1.529e+08	4.457e+07	<2e-16	***
Ticket19947	5.915e+15	8.958e+07	6.603e+07	<2e-16	***
Ticket19950	1.383e+16	3.400e+08	4.069e+07	<2e-16	***
Ticket19952	1.922e+15	8.878e+07	2.165e+07	<2e-16	***
Ticket19988	3.666e+15	7.592e+07	4.829e+07	<2e-16	***
Ticket19996	-1.724e+15	7.957e+07	-2.167e+07	<2e-16	***
Ticket2003	9.034e+15	1.852e+08	4.879e+07	<2e-16	***
Ticket211536	1.119e+15	9.848e+07	1.136e+07	<2e-16	***
Ticket21440	8.771e+15	2.392e+08	3.666e+07	<2e-16	***
Ticket218629	1.346e+15	9.992e+07	1.347e+07	<2e-16	***
Ticket219533	-7.691e+15	2.073e+08	-3.710e+07	<2e-16	***
Ticket220367	1.119e+15	9.848e+07	1.136e+07	<2e-16	***
Ticket220845	1.148e+16	2.342e+08	4.904e+07	<2e-16	***
Ticket2223	8.822e+15	2.438e+08	3.619e+07	<2e-16	***
Ticket223596	3.277e+15	1.013e+08	3.233e+07	<2e-16	***
Ticket226593	-6.316e+15	2.047e+08	-3.085e+07	<2e-16	***
Ticket226875	9.049e+15	1.842e+08	4.911e+07	<2e-16	***
Ticket228414	9.036e+15	1.851e+08	4.883e+07	<2e-16	***
Ticket229236	1.136e+15	9.819e+07	1.157e+07	<2e-16	***
Ticket230080	5.219e+14	9.305e+07	5.609e+06	<2e-16	***
Ticket230136	5.665e+15	1.211e+08	4.677e+07	<2e-16	***
Ticket230433	3.034e+15	8.672e+07	3.499e+07	<2e-16	***
Ticket231919	1.865e+15	9.777e+07	1.908e+07	<2e-16	***
Ticket231945	4.276e+14	9.599e+07	4.454e+06	<2e-16	***
Ticket233639	1.112e+15	9.873e+07	1.126e+07	<2e-16	***
Ticket233866	1.113e+15	9.866e+07	1.129e+07	<2e-16	***
Ticket234360	1.158e+15	9.838e+07	1.177e+07	<2e-16	***
Ticket234604	3.028e+15	9.983e+07	3.033e+07	<2e-16	***
Ticket234686	1.115e+15	9.859e+07	1.131e+07	<2e-16	***
Ticket234818	-6.299e+15	2.059e+08	-3.059e+07	<2e-16	***
Ticket236171	1.102e+15	9.913e+07	1.112e+07	<2e-16	***
Ticket236852	3.046e+15	9.981e+07	3.052e+07	<2e-16	***
Ticket236853	7.129e+15	1.833e+08	3.890e+07	<2e-16	***
Ticket237442	1.403e+15	1.004e+08	1.398e+07	<2e-16	***
Ticket237565	2.056e+15	1.051e+08	1.955e+07	<2e-16	***
Ticket237671	-1.923e+15	9.981e+07	-1.926e+07	<2e-16	***
Ticket237736	-5.025e+14	1.130e+08	-4.448e+06	<2e-16	***
Ticket237789	1.037e+16	2.139e+08	4.846e+07	<2e-16	***
Ticket237798	5.650e+15	9.821e+07	5.754e+07	<2e-16	***

Ticket239853	1.125e+15	8.166e+07	1.377e+07	<2e-16	***
Ticket239854	-3.378e+15	9.834e+07	-3.435e+07	<2e-16	***
Ticket239855	1.125e+15	9.834e+07	1.143e+07	<2e-16	***
Ticket239856	1.125e+15	9.834e+07	1.143e+07	<2e-16	***
Ticket239865	1.099e+15	9.933e+07	1.106e+07	<2e-16	***
Ticket240929	2.863e+15	9.900e+07	2.892e+07	<2e-16	***
Ticket24160	1.630e+16	4.074e+08	4.002e+07	<2e-16	***
Ticket243847	3.007e+14	8.819e+07	3.410e+06	<2e-16	***
Ticket244252	-3.561e+15	8.655e+07	-4.115e+07	<2e-16	***
Ticket244270	5.630e+15	9.831e+07	5.727e+07	<2e-16	***
Ticket244278	-1.295e+13	9.502e+07	-1.363e+05	<2e-16	***
Ticket244310	1.147e+15	9.821e+07	1.168e+07	<2e-16	***
Ticket244367	1.298e+15	9.790e+07	1.326e+07	<2e-16	***
Ticket244373	5.628e+15	9.834e+07	5.723e+07	<2e-16	***
Ticket248698	5.636e+15	9.822e+07	5.737e+07	<2e-16	***
Ticket248706	4.420e+15	1.120e+08	3.945e+07	<2e-16	***
Ticket248723	1.141e+15	9.818e+07	1.162e+07	<2e-16	***
Ticket248727	-3.394e+14	7.905e+07	-4.294e+06	<2e-16	***
Ticket248731	1.390e+15	9.989e+07	1.392e+07	<2e-16	***
Ticket248738	2.972e+15	1.068e+08	2.783e+07	<2e-16	***
Ticket248740	1.121e+15	9.843e+07	1.139e+07	<2e-16	***
Ticket248747	-1.948e+15	9.925e+07	-1.963e+07	<2e-16	***
Ticket250643	1.162e+15	9.848e+07	1.179e+07	<2e-16	***
Ticket250644	-9.281e+14	8.522e+07	-1.089e+07	<2e-16	***
Ticket250646	1.125e+15	9.834e+07	1.143e+07	<2e-16	***
Ticket250647	-4.027e+15	9.322e+07	-4.320e+07	<2e-16	***
Ticket250648	2.566e+15	9.938e+07	2.582e+07	<2e-16	***
Ticket250649	-2.102e+15	9.663e+07	-2.175e+07	<2e-16	***
Ticket250651	4.523e+15	1.818e+08	2.487e+07	<2e-16	***
Ticket250652	2.536e+15	1.018e+08	2.490e+07	<2e-16	***
Ticket250653	-3.367e+15	9.834e+07	-3.424e+07	<2e-16	***
Ticket250655	-4.091e+14	8.620e+07	-4.746e+06	<2e-16	***
Ticket2620	1.244e+16	2.428e+08	5.121e+07	<2e-16	***
Ticket2625	9.704e+15	2.506e+08	3.872e+07	<2e-16	***
Ticket2626	9.857e+15	2.440e+08	4.039e+07	<2e-16	***
Ticket2627	4.154e+15	2.376e+08	1.748e+07	<2e-16	***
Ticket2628	7.975e+15	2.411e+08	3.307e+07	<2e-16	***
Ticket26360	2.677e+15	1.016e+08	2.634e+07	<2e-16	***
Ticket2641	7.949e+15	2.422e+08	3.282e+07	<2e-16	***
Ticket2647	7.947e+15	2.421e+08	3.282e+07	<2e-16	***
Ticket2648	6.483e+15	2.200e+08	2.947e+07	<2e-16	***
Ticket2649	9.856e+15	2.440e+08	4.039e+07	<2e-16	***
Ticket2650	1.344e+16	3.030e+08	4.438e+07	<2e-16	***
Ticket2651	6.447e+15	2.238e+08	2.881e+07	<2e-16	***
Ticket2653	9.617e+15	2.443e+08	3.937e+07	<2e-16	***
Ticket2659	6.796e+15	2.368e+08	2.869e+07	<2e-16	***
Ticket2661	9.474e+15	2.417e+08	3.919e+07	<2e-16	***
Ticket2662	1.478e+16	3.590e+08	4.117e+07	<2e-16	***
Ticket2664	7.947e+15	2.421e+08	3.282e+07	<2e-16	***
Ticket2665	2.778e+15	2.356e+08	1.179e+07	<2e-16	***
Ticket2666	5.630e+15	2.150e+08	2.619e+07	<2e-16	***
Ticket2667	9.361e+15	2.427e+08	3.857e+07	<2e-16	***
Ticket2668	9.218e+15	2.639e+08	3.492e+07	<2e-16	***
Ticket2669	7.934e+15	2.429e+08	3.267e+07	<2e-16	***
Ticket26707	7.164e+15	1.811e+08	3.956e+07	<2e-16	***
Ticket2671	7.949e+15	2.422e+08	3.282e+07	<2e-16	***
Ticket2672	7.937e+15	2.426e+08	3.272e+07	<2e-16	***

```

Ticket2674      7.947e+15  2.421e+08  3.282e+07  <2e-16 ***
Ticket2678      3.103e+15  2.368e+08  1.311e+07  <2e-16 ***
Ticket2680      1.136e+16  2.982e+08  3.808e+07  <2e-16 ***
Ticket2683      7.601e+15  2.360e+08  3.221e+07  <2e-16 ***
Ticket2685      7.949e+15  2.422e+08  3.282e+07  <2e-16 ***
Ticket2686      7.949e+15  2.422e+08  3.282e+07  <2e-16 ***
Ticket2687      9.359e+15  2.429e+08  3.853e+07  <2e-16 ***
Ticket2689      8.772e+15  2.997e+08  2.927e+07  <2e-16 ***
Ticket2690      8.087e+15  2.386e+08  3.389e+07  <2e-16 ***
Ticket2691      2.862e+15  2.352e+08  1.217e+07  <2e-16 ***
Ticket2693      7.937e+15  2.427e+08  3.270e+07  <2e-16 ***
Ticket2694      7.947e+15  2.421e+08  3.282e+07  <2e-16 ***
Ticket2695      8.083e+15  2.388e+08  3.385e+07  <2e-16 ***
Ticket2697      7.946e+15  2.423e+08  3.279e+07  <2e-16 ***
Ticket2699      8.901e+15  2.556e+08  3.482e+07  <2e-16 ***
Ticket2700      7.949e+15  2.422e+08  3.282e+07  <2e-16 ***
Ticket27042     3.533e+15  8.017e+07  4.407e+07  <2e-16 ***
Ticket27267     2.571e+15  9.948e+07  2.584e+07  <2e-16 ***
[ reached getOption("max.print") -- omitted 375 rows ]
---

```

```

Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

(Dispersion parameter for binomial family taken to be 1)

```

```

Null deviance: 943.08  on 711  degrees of freedom
Residual deviance: 576.70  on 141  degrees of freedom
(1 observation deleted due to missingness)
AIC: 1718.7

```

```

Number of Fisher Scoring iterations: 16

```

```

Call:
glm(formula = Survived ~ Pclass + Age + SibSp + Parch + Title +
     I(Ticket > 2), family = binomial, data = d[inTrain, ])

```

```

Deviance Residuals:

```

```

    Min       1Q   Median       3Q      Max
-2.6268 -0.5138 -0.3596  0.5342  2.6923

```

```

Coefficients:

```

```

              Estimate Std. Error z value Pr(>|z|)
(Intercept)    5.69803    0.72768   7.830 4.86e-15 ***
Pclass         -1.09370    0.16973  -6.444 1.16e-10 ***
Age            -0.02927    0.01073  -2.726 0.006403 **
SibSp          -0.48954    0.13350  -3.667 0.000246 ***
Parch          -0.26606    0.14547  -1.829 0.067404 .
TitleMiss      -0.31742    0.54436  -0.583 0.559818
TitleMr        -3.33644    0.59484  -5.609 2.04e-08 ***
TitleMrs        0.56676    0.61818   0.917 0.359237
I(Ticket > 2)TRUE -0.82075    0.34158  -2.403 0.016271 *

```

```

---

```

```

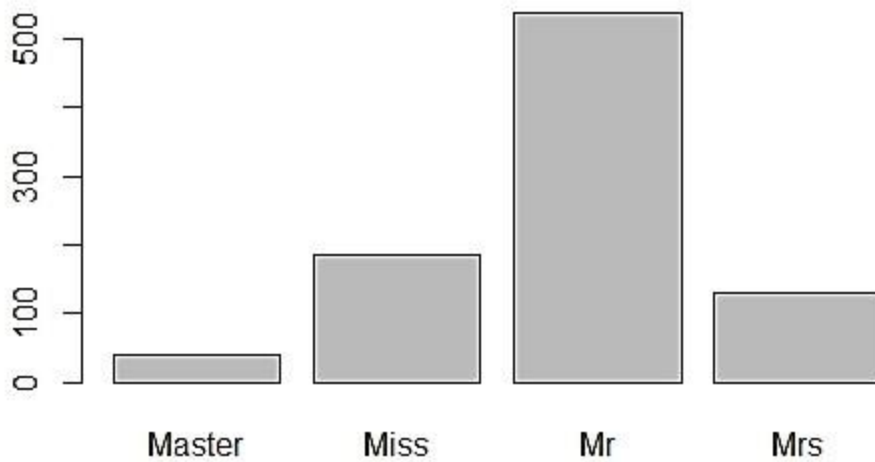
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 945.03 on 712 degrees of freedom  
Residual deviance: 567.25 on 704 degrees of freedom  
AIC: 585.25

Number of Fisher Scoring iterations: 5



```
p1 <- ggplot(data=train,aes(x=Age)) + geom_histogram(aes(fill=Survived),bins = 40) +  
coord_flip()
```

```
p2 <- ggplot(data=train,aes(x=Fare)) + geom_histogram(aes(fill=Survived),bins = 40) +  
coord_flip()
```

```
grid.arrange(p1,p2,nrow=1)
```

```
summary(train$Fare)
```



```

get_legend<-function(myggplot){
  tmp <- ggplot_gtable(ggplot_build(myggplot))
  leg <- which(sapply(train, function(x) x$name) == "guide-box")
  legend <- tmp$grobs[[leg]]
  return(legend)
}

```

```

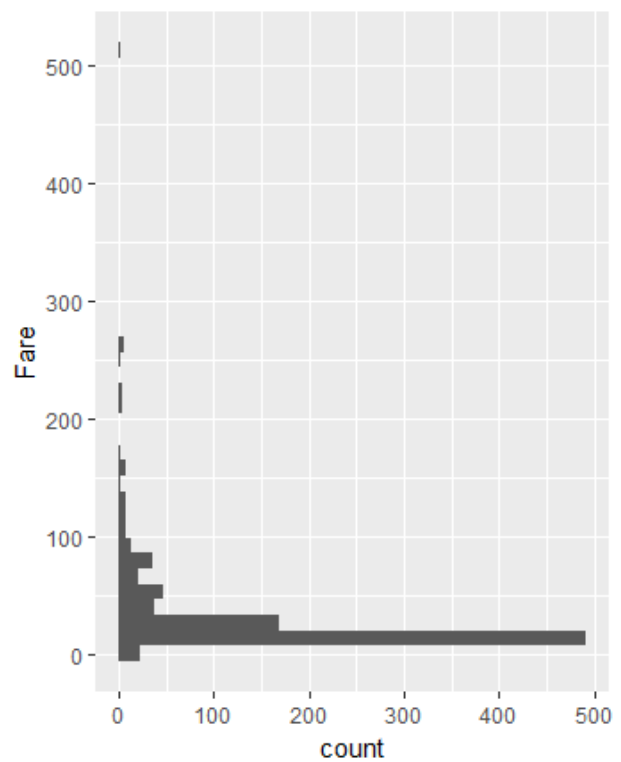
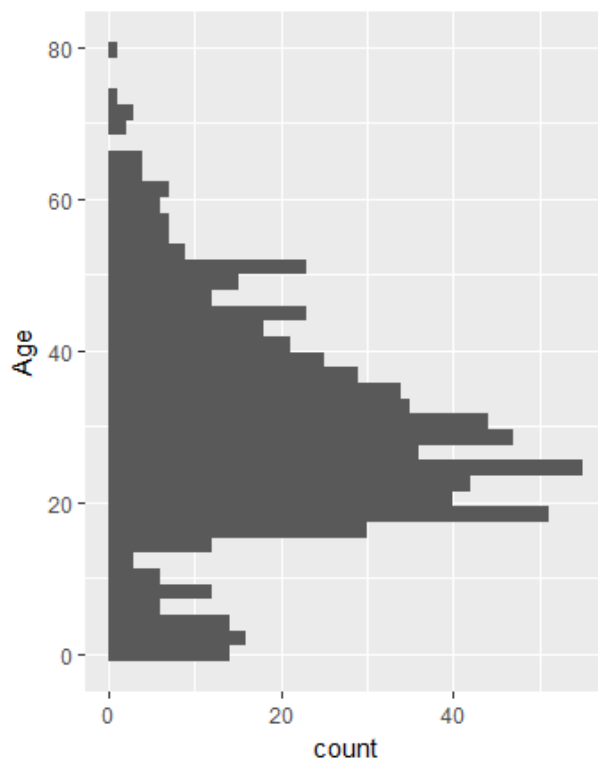
p <- lapply(X = c('Pclass','Sex','SibSp','Parch','Embarked'),

```

```

  FUN = function(x) ggplot(data = train)+
    aes_string(x=x,fill='Survived')+
    geom_bar(position="dodge")+
    theme(legend.position="none"))

```



```

``{r}

summary(train$Embarked)

train.imp <- train
train.imp$Embarked[is.na(train.imp$Embarked)] <- 'S'
train.imp$title <- str_extract(pattern = '[a-zA-Z]+(?:\\.|\\.)',string = train.imp$Name)
train.imp$title <- as.factor(train.imp$title)
ggplot(train.imp,aes(x=title,y=Age))+
  geom_jitter(shape=21,alpha=.6,col='blue')+
  stat_summary(aes(y = Age,group=1), fun.y=median, colour="red",
geom="point",group=1)+
  theme_bw()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1),legend.position="none")+
  labs(caption='red points are median values')

train.imp$title <- as.character(train.imp$title)
train.imp$title[train.imp$title %in% c('Capt','Col','Major')] <- 'Officer'
train.imp$title[train.imp$title %in%
c('Don','Dr','Rev','Sir','Jonkheer','Countess','Lady','Dona')] <- 'Royalty'
train.imp$title[train.imp$title %in% c('Mrs','Mme')] <- 'Mrs'
train.imp$title[train.imp$title %in% c('Ms','Mlle')] <- 'Miss'
train.imp$title <- as.factor(train.imp$title)
ggplot(train.imp,aes(x=title,y=Age))+
  geom_jitter(color='blue',shape=21,alpha=.7)+
  stat_summary(aes(y = Age,group=1), fun.y=median, colour="red",
geom="point",group=1)+
  theme_bw()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))+
  labs(caption='red points are median values')

```

```

age.predictors <- train.imp %>%
  dplyr::select(-Survived,-Cabin,-Ticket,-Name) %>%
  dplyr::filter(complete.cases(.))
ctrl <- trainControl(method = "repeatedcv",
  repeats = 5)
rpartGrid <- data.frame(maxdepth = seq(2,10,1))
rpartFit_ageimputation <- train(x=age.predictors[,-3],
  y=age.predictors$Age,
  method='rpart2',
  trControl = ctrl,
  tuneGrid = rpartGrid
)
rpartFit_ageimputation
## CART
##
## 508 samples
## 7 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 457, 457, 457, 457, 457, 457, ...
## Resampling results across tuning parameters:
##
## maxdepth RMSE  Rsquared  MAE
## 2    12.02414 0.3171031 9.443687
## 3    11.30498 0.3985131 8.707856
## 4    11.42463 0.3882499 8.782511
## 5    11.27085 0.4038018 8.639549

```

```

## 6    11.39825 0.3930011 8.720958
## 7    11.43177 0.3890118 8.744528
## 8    11.47797 0.3851413 8.783542
## 9    11.48005 0.3848860 8.783870
## 10   11.48005 0.3848860 8.783870
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was maxdepth = 5.
plot(rpartFit_ageimputation)
rpart.plot::rpart.plot(rpartFit_ageimputation$finalModel, extra=101, box.palette="GnBu")
save(rpartFit_ageimputation,file = 'rpartFit_ageimputation')
missing_age <- is.na(train.imp$Age)
age.predicted <- predict(rpartFit_ageimputation, newdata = train.imp[missing_age,])
train.imp[missing_age,'Age'] <- age.predicted

train.imp %>%
  mutate(Age_Imputed = missing_age) %>%
  ggplot(aes(x=title,y=Age))+
  stat_summary(aes(y = Age,group=1), fun.y=median, colour="red",
geom="point",group=1)+
  geom_jitter(aes(y=Age,col=Age_Imputed,shape=Age_Imputed))+
  theme_bw()+
  theme(axis.text.x = element_text(angle = 45, hjust = 1),legend.position="none")+
  labs(caption='green points are imputed values')
train.imp$child <- 0
train.imp$child[train.imp$Age<18] <- 1
train.imp$Seniors <- ifelse(train.imp$Age>60,1,0)
train.imp$TotalFam <- train.imp$SibSp + train.imp$Parch + 1

```

```

train.imp$LargeFamily <- ifelse(train.imp$TotalFam>4,1,0)
train.imp$Name <- NULL
train.imp$CabinCode <- map_chr(train$Cabin,~str_split(string = .x,pattern = "")[[1]][1])
train.imp$CabinCode[is.na(train.imp$CabinCode)] <- 'U'
train.imp$CabinCode <- as.factor(train.imp$CabinCode)

train.imp$CabinNum <- as.numeric(map_chr(train$Cabin,~str_split(string = .x,pattern =
'[a-zA-Z]')[[1]][2]))
train.imp$CabinNum <- map_int(train.imp$CabinNum, ~as.integer(str_split(.x,pattern =
",simplify = T)[1][1]))
train.imp$CabinNum[is.na(train.imp$CabinNum)] <- 0

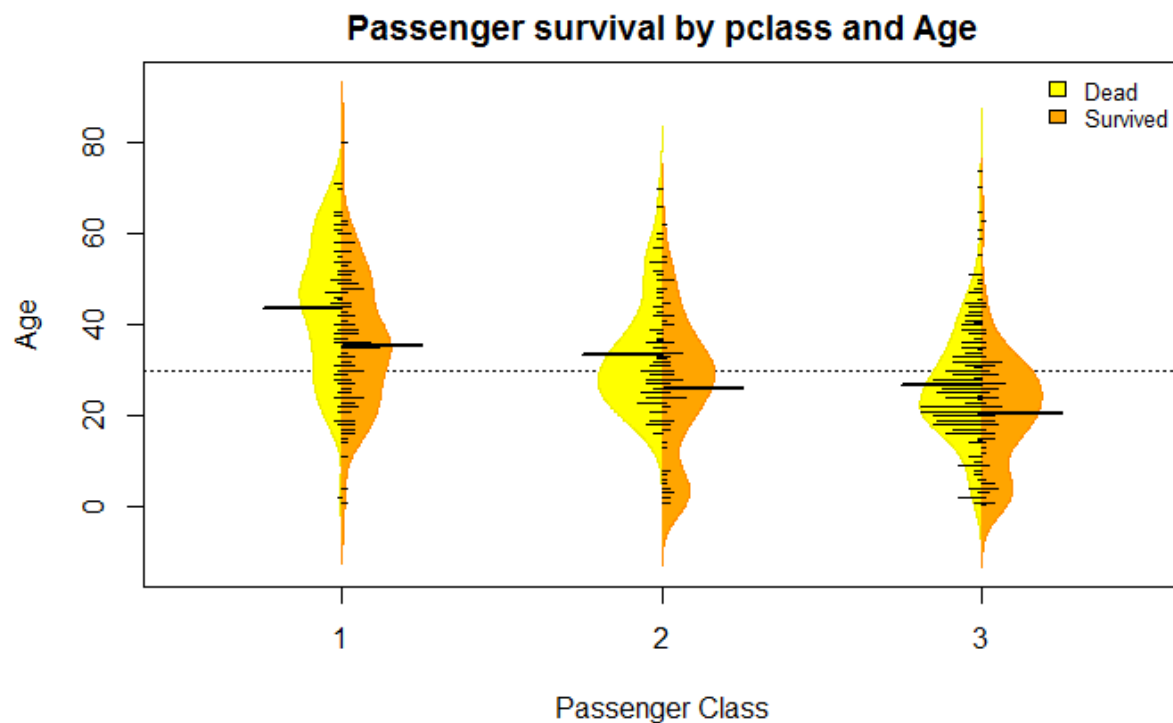
train.imp$TopDeck <- ifelse(train.imp$CabinCode %in% c('A','B'),1,0)
train.imp$MidDeck <- ifelse(train.imp$CabinCode %in% c('C','D'),1,0)
train.imp$LowerDeck <- ifelse(train.imp$TopDeck==0 & train.imp$MidDeck==0 ,1,0)

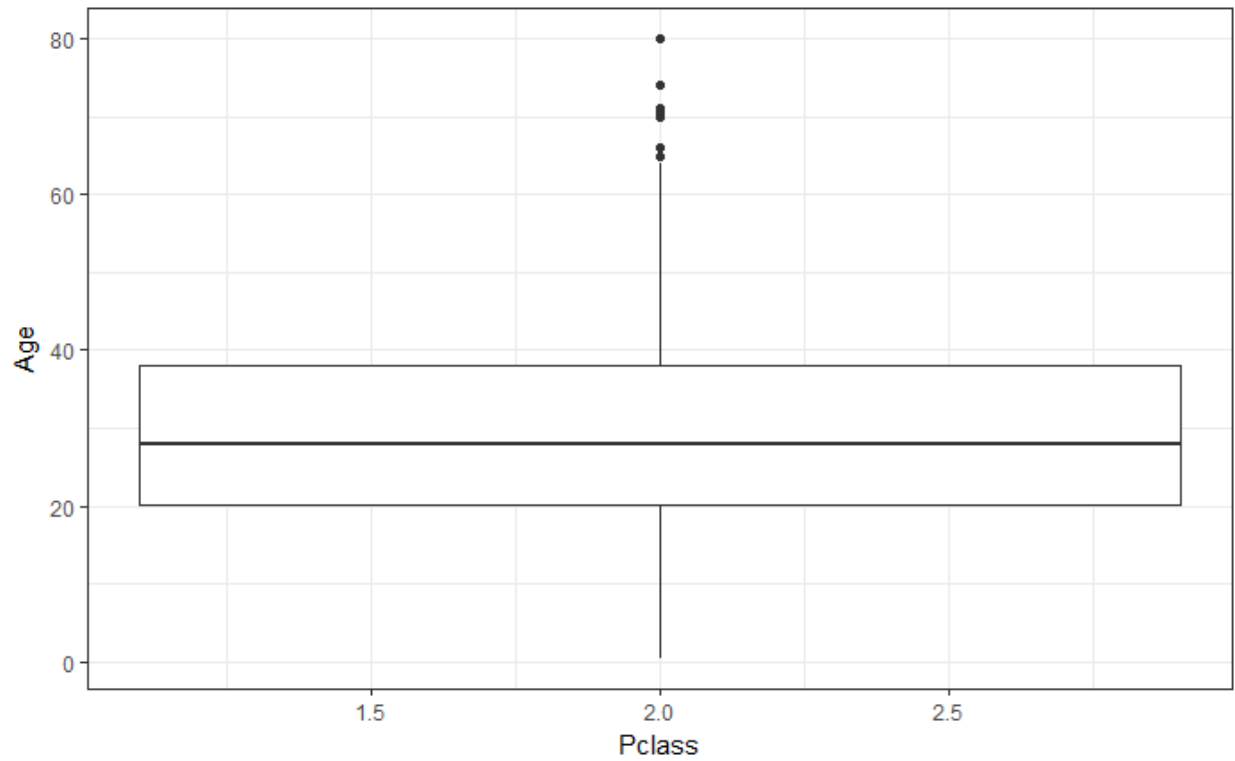
train.imp$NumberofCabins <- map_int(train$Cabin,~str_split(string = .x,pattern = '')[[1]]
%>% length)
train.imp$Cabin <- NULL
train.imp$Ticket %>% table() %>% as.numeric() %>% table()
## .
## 1 2 3 4 5 6 7
## 430 60 15 3 1 1 1
train.imp %>% group_by(Pclass) %>% dplyr::select(Ticket,Pclass) %>% sample_n(5)

...

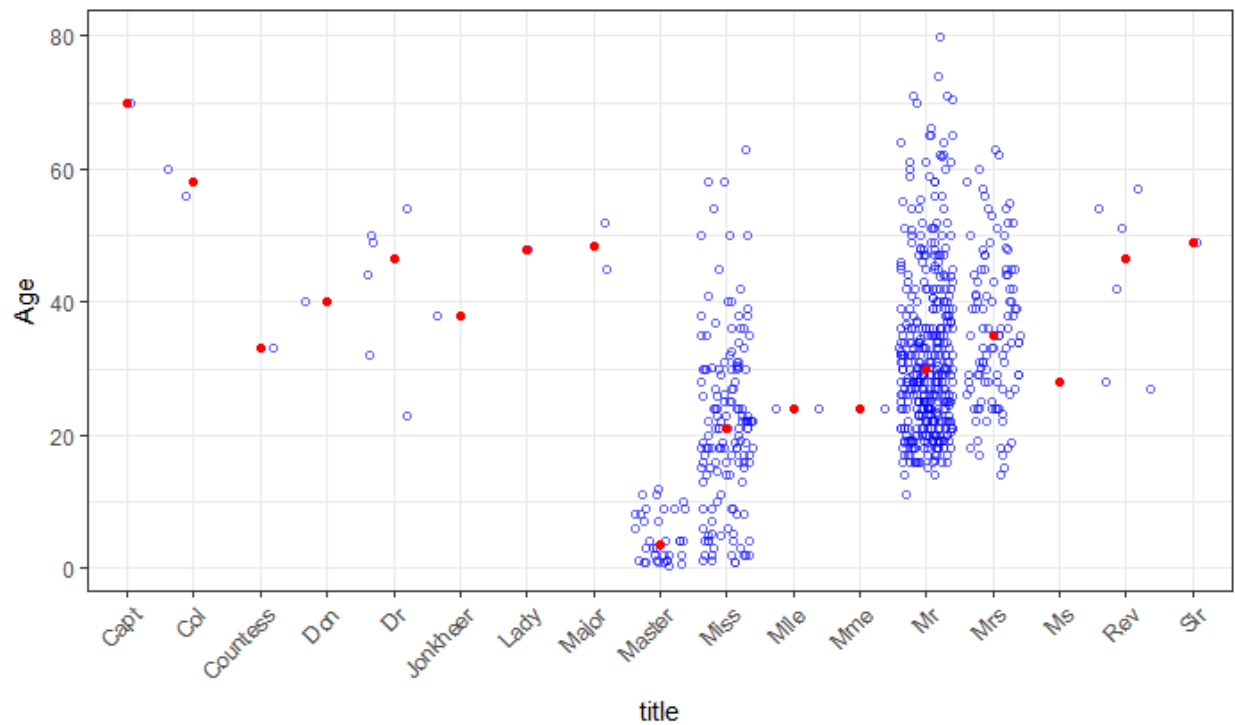
```

```
ggplot(train,aes(y=Age,x=Pclass))+geom_boxplot(aes(fill=Survived))+theme_bw()
warning messages:
1: Continuous x aesthetic -- did you forget aes(group=...)?
2: Removed 177 rows containing non-finite values (stat_boxplot).
> beanplot(Age~Survived*Pclass,side='b',train,col=list('yellow','orange'),
+         border = c('yellow2','darkorange'),ll = 0.05,boxwex = .5,
+         main='Passenger survival by pclass and Age',xlab='Passenger Class'
,ylab='Age')
> legend('topright', fill = c('yellow','orange'), legend = c("Dead", "Survive
d"),bty = 'n',cex = .8)
```





```
stat_summary(aes(y = Age, group=1), fun.y=median, colour="red", geom="point", group=1)+
theme_bw()+ theme(axis.text.x = element_text(angle = 45, hjust =
1), legend.position="none")+ labs(caption='red points are median values')
```

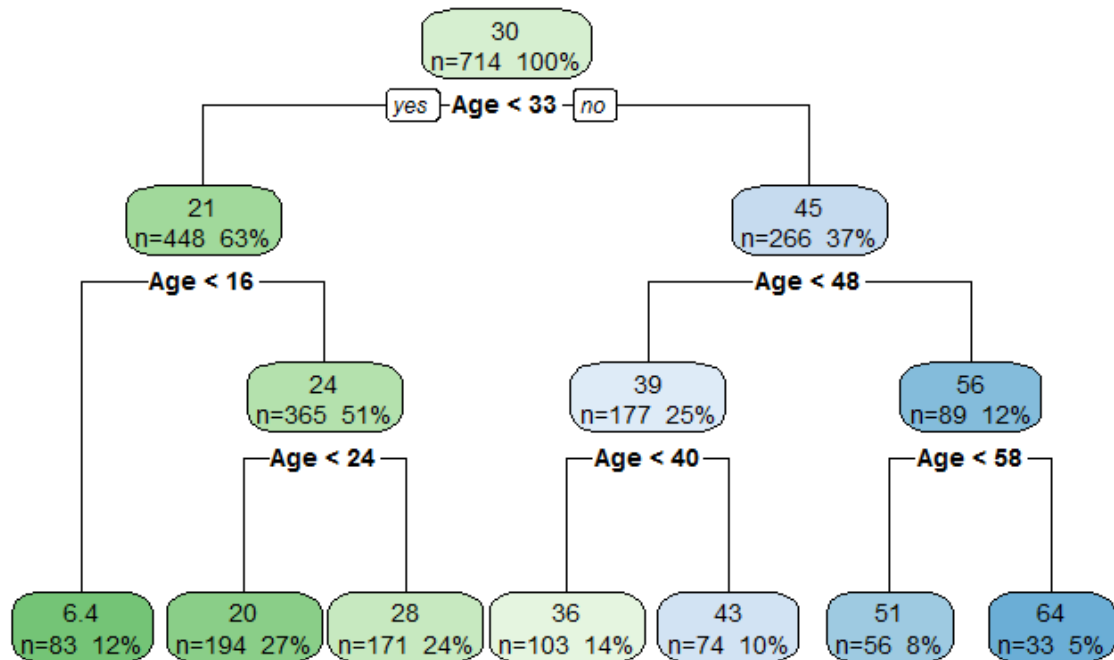


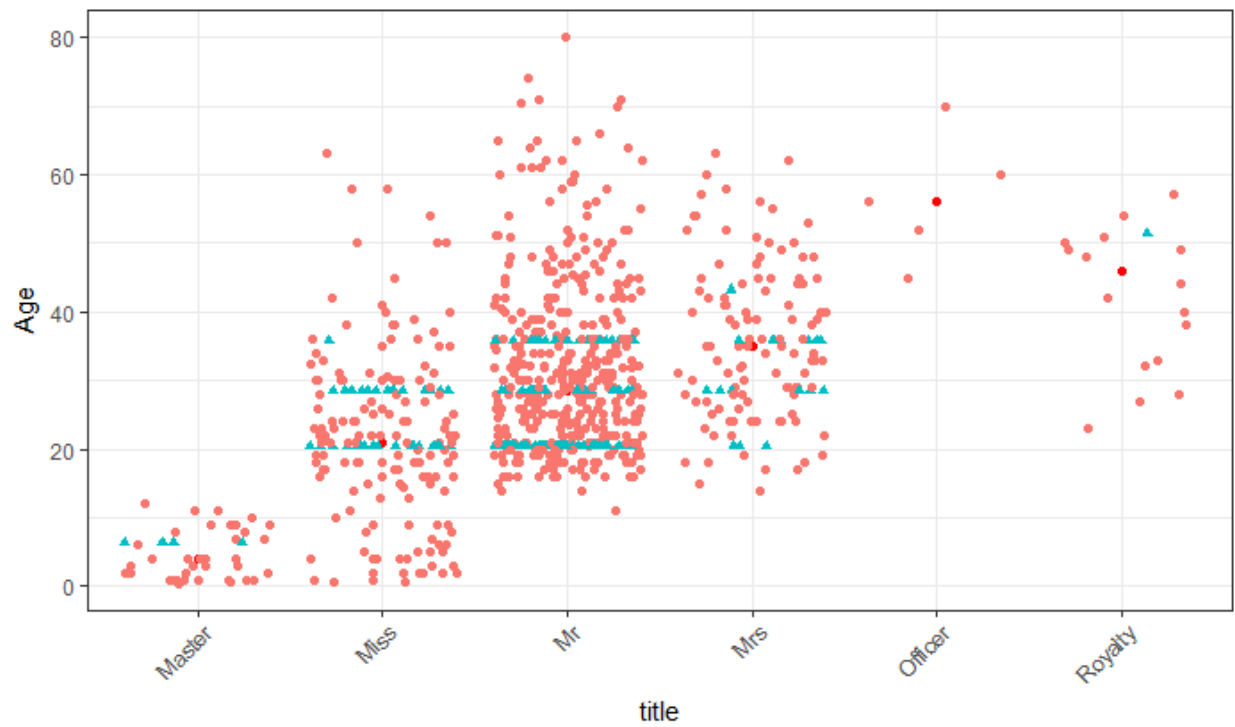
red points are median values

```
## CART
##
## 508 samples
## 7 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 457, 457, 457, 457, 457, 457, ...
## Resampling results across tuning parameters:
##
##   maxdepth  RMSE      Rsquared  MAE
##   2         12.02414  0.3171031  9.443687
##   3         11.30498  0.3985131  8.707856
##   4         11.42463  0.3882499  8.782511
##   5         11.27085  0.4038018  8.639549
##   6         11.39825  0.3930011  8.720958
##   7         11.43177  0.3890118  8.744528
```



```
##      8      11.47797  0.3851413  8.783542
##      9      11.48005  0.3848860  8.783870
##     10      11.48005  0.3848860  8.783870
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was maxdepth = 5.
plot(rpartFit_ageimputation)
rpart.plot::rpart.plot(rpartFit_ageimputation$finalModel, extra=101, box.pal
tte="GnBu")
save(rpartFit_ageimputation,file = 'rpartFit_ageimputation')
```





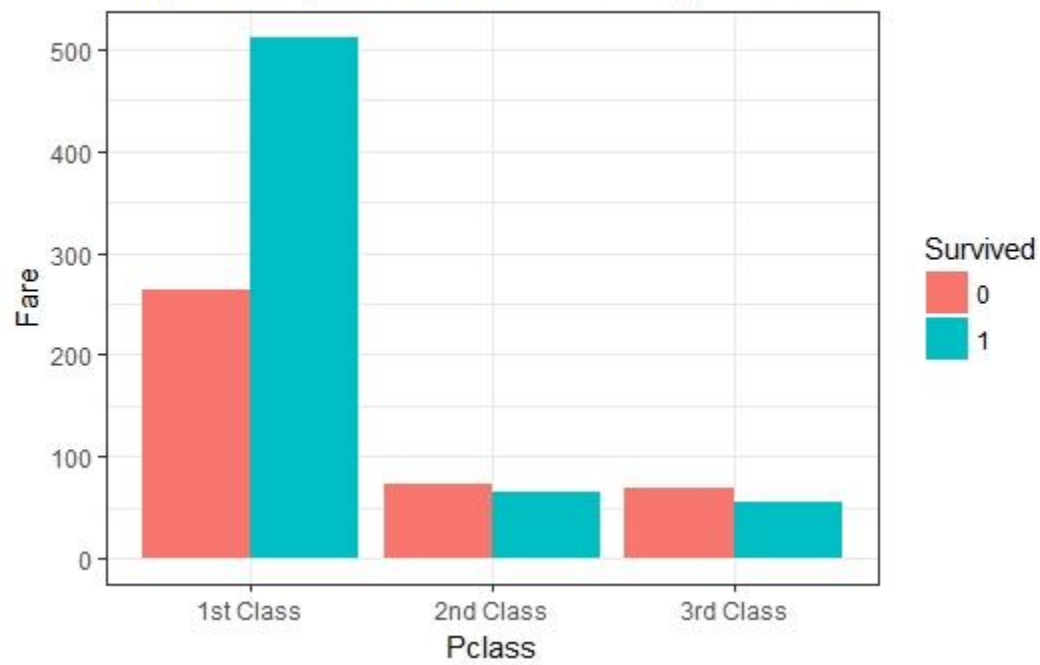
green points are imputed values

Ticket	Pclass
<chr>	<int>
113767	1
17421	1
PC 17582	1
113510	1
13507	1
S.O.C. 14879	2
244373	2
239853	2
C.A. 31921	2
236853	2
Next	
12	
Previous	
1-10 of 15 rows	
Ticket	Pclass
<chr>	<int>
2665	3
2691	3
A/4 48871	3
349204	3
349248	3

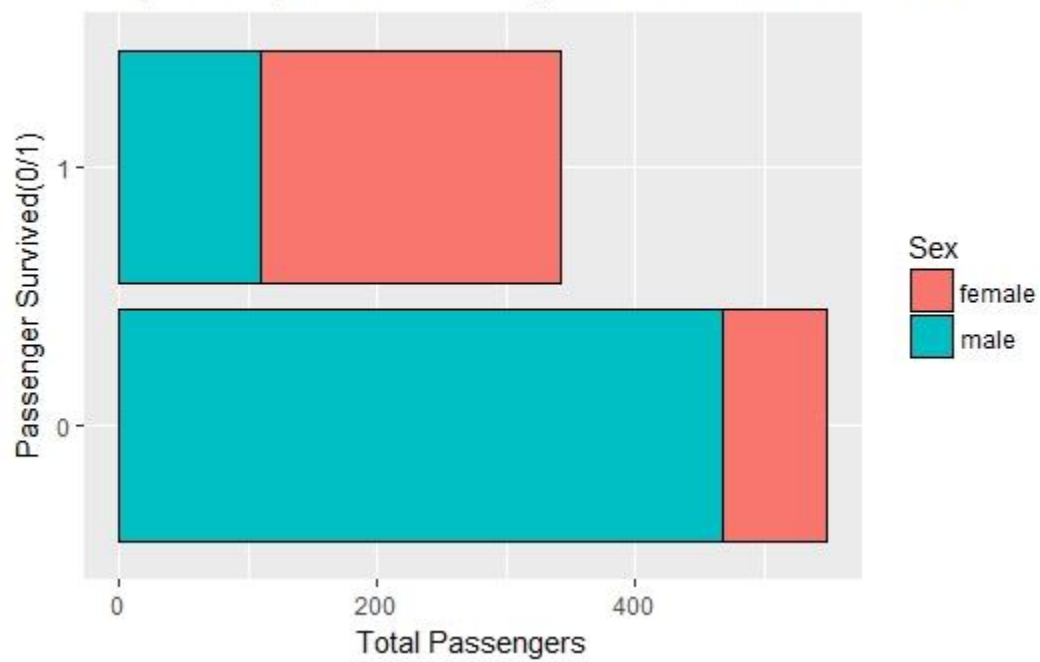
**b. Represent the proportion of people survived from the family size using a graph.**

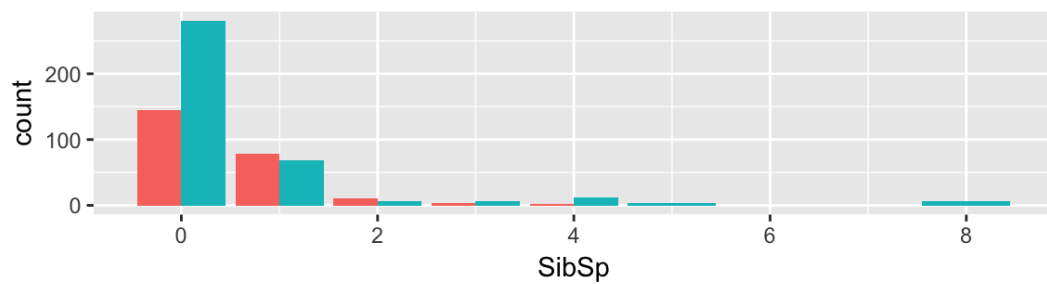
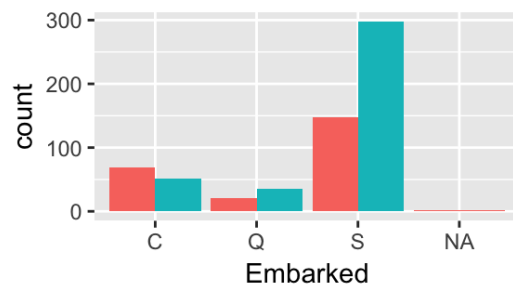
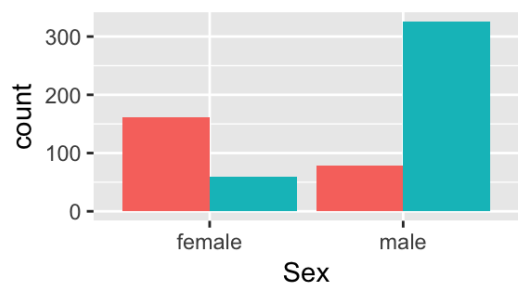
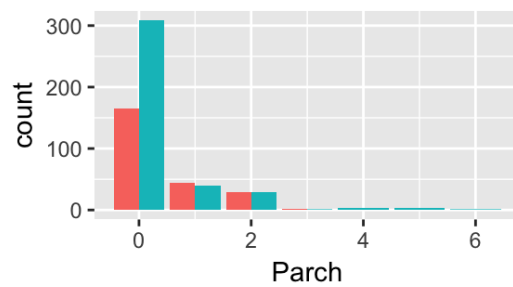
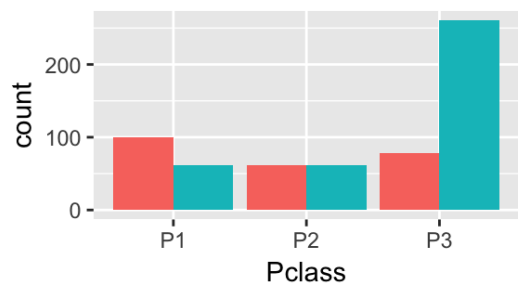
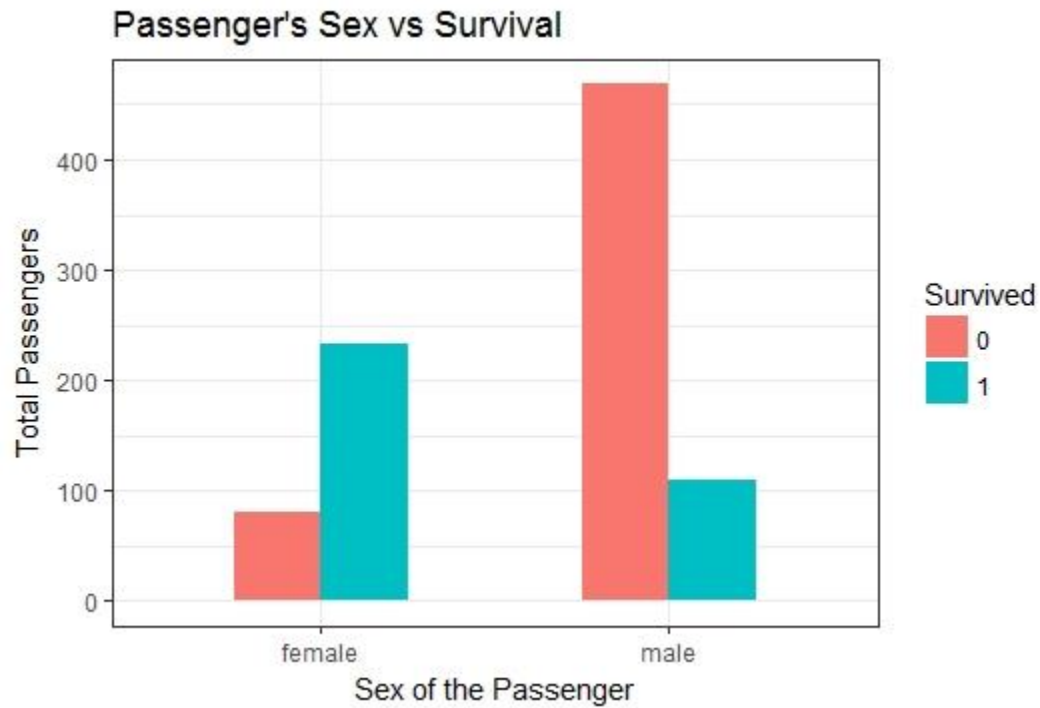
```
get_legend<-function(myggplot){  
  tmp <- ggplot_gtable(ggplot_build(myggplot))  
  leg <- which(sapply(tmp$grobs,function(x) x$name) == "guide-box")  
  legend <- tmp$grobs[[leg]]  
  return(legend)  
}  
p <- lapply(X = c('Pclass','Sex','SibSp','Parch','Embarked'),  
  FUN = function(x) ggplot(data = train)+  
    aes_string(x=x,fill='Survived')+  
    geom_bar(position="dodge")+  
    theme(legend.position="none"))  
legend <- get_legend(ggplot(data = train,aes(x=Pclass,fill=Survived))+geom_bar())  
grid.arrange(p[[1]],p[[2]],p[[3]],p[[4]],p[[5]], legend,layout_matrix = cbind(c(1,2,3),  
c(4,5,3), c(6,6,6)), widths=c(3,3,1))
```

Barplot to represent Fare vs Passenger Class

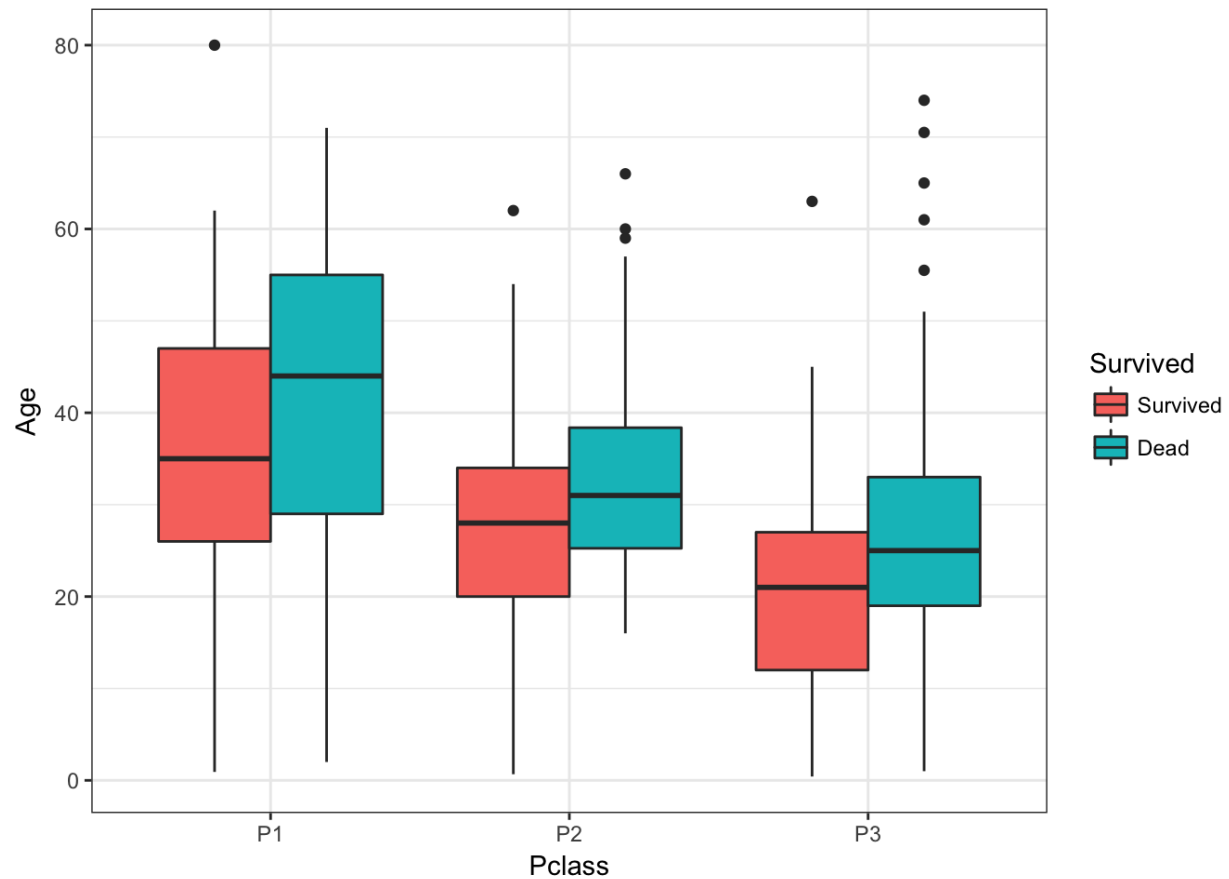


Barplot to represent Passenger Count who Survived vs who D





```
ggplot(train,aes(y=Age,x=Pclass))+geom_boxplot(aes(fill=Survived))+theme_bw()
> beanplot(Age~Survived*Pclass,side='b',train,col=list('yellow','orange'),
+         border = c('yellow2','darkorange'),ll = 0.05,boxwex = .5,
+         main='Passenger survival by pclass and Age',xlab='Passenger Class'
+         ,ylab='Age')
```



c. Impute the missing values in Age variable using Mice Library, create two different graphs showing Age distribution before and after imputation

```
summary(training)
  PassengerId      Survived      Pclass         Name         Sex
Min.   : 1.0      Min.   :0.0000   Min.   :1.000   Length:891   Length:8
91
1st Qu.:223.5    1st Qu.:0.0000   1st Qu.:2.000   Class :character   Class :c
character
Median :446.0    Median :0.0000   Median :3.000   Mode  :character   Mode  :c
character
Mean   :446.0    Mean   :0.3838   Mean   :2.309
3rd Qu.:668.5    3rd Qu.:1.0000   3rd Qu.:3.000
Max.   :891.0    Max.   :1.0000   Max.   :3.000

  Age      SibSp      Parch      Ticket      Fare
Min.   : 0.42   Min.   :0.000   Min.   :0.0000   Length:891   Min.   :
0.00
1st Qu.:20.12   1st Qu.:0.000   1st Qu.:0.0000   Class :character   1st Qu.:
7.91
Median :28.00   Median :0.000   Median :0.0000   Mode  :character   Median :
14.45
Mean   :29.70   Mean   :0.523   Mean   :0.3816
32.20
3rd Qu.:38.00   3rd Qu.:1.000   3rd Qu.:0.0000
31.00
Max.   :80.00   Max.   :8.000   Max.   :6.0000
512.33
NA's   :177
  Cabin      Embarked
Length:891   Length:891
Class :character   Class :character
Mode  :character   Mode  :character
dim(training)
[1] 891 12
> str(training)
Classes 'tbl_df', 'tbl' and '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 ...
 $ Name       : chr  "Braund, Mr. Owen Harris" "Cumings, Mrs. John Bradley (Florence Briggs Thayer)" "Heikkinen, Miss. Laina" "Futrelle, Mrs. Jacques Heath (Lily May Peel)" ...
 $ Sex        : chr  "male" "female" "female" "female" ...
 $ Age        : num  22 38 26 35 35 NA 54 2 27 14 ...
 $ SibSp      : int  1 1 0 1 0 0 0 3 0 1 ...
 $ Parch      : int  0 0 0 0 0 0 0 1 2 0 ...
 $ Ticket     : chr  "A/5 21171" "PC 17599" "STON/O2. 3101282" "113803" ...
 $ Fare       : num  7.25 71.28 7.92 53.1 8.05 ...
 $ Cabin      : chr  NA "C85" NA "C123" ...
 $ Embarked   : chr  "S" "C" "S" "S" ...
```

```

training[training==""] <- NA
> a <- apply(training,2,is.na)
> summary(a)
 PassengerId      Survived      Pclass         Name         Sex
Mode :logical    Mode :logical Mode :logical Mode :logical Mode :logical
1
FALSE:891        FALSE:891        FALSE:891        FALSE:891        FALSE:891

  Age      SibSp      Parch      Ticket      Fare
Mode :logical Mode :logical Mode :logical Mode :logical Mode :logical
1
FALSE:714        FALSE:891        FALSE:891        FALSE:891        FALSE:891
TRUE :177

  Cabin      Embarked
Mode :logical Mode :logical
FALSE:204      FALSE:889
TRUE :687      TRUE :2
apply(a,2,sum)
 PassengerId      Survived      Pclass         Name         Sex         Age
 SibSp
0           0           0           0           0           0          177

  Parch      Ticket      Fare      Cabin      Embarked
0           0           0           687           2

```

It can be seen that Age,Cabin and Embarked variables have missing values. Cabin has most number of missing values. These missing values can be found by using 'Multivariate Imputation by Chained Equations (MICE)' package

```

training$Salutation <- gsub('(.*, )|(\\.*)', '',training$Name)
> table(training$Sex,training$Salutation)

      Capt Col Don  Dr Jonkheer Lady Major Master Miss Mlle Mme  Mr Mrs  M
s Rev Sir
female    0  0  0  1      0  1  0      0 182  2  1  0 125
1  0  0
male      1  2  1  6      1  0  2  40  0  0  0 517  0
0  6  1

      the Countess
female      1
male      0
misc <- c("Capt","Col","Don","Dr","Jonkheer","Lady","Major","Rev","Sir","the
Countess","Dona")
> training$Salutation[training$Salutation == "Mlle"] <- "Miss"
> training$Salutation[training$Salutation == "Mme"] <- "Miss"
> training$Salutation[training$Salutation %in% misc] <- "Misc"
> table(training$Sex,training$Salutation)

      Master Misc Miss  Mr Mrs  Ms
female    0  3 185  0 125  1
male     40 20  0 517  0  0

```

```

training$Surname <- sapply(training$Name,function(x) strsplit(x, split = '[,.]')[[1]][1])
> s <- nlevels(factor(training$Surname))
> paste('we have', s, 'unique surnames in the training dataset amongst',nrow(training), 'passange
rs.')
```



```
[1] "We have 667 unique surnames in the training dataset amongst 891 passangers."
```

```
>
```

```
training$Deck <- substr(training$Cabin,1,1)
> paste("Titanic has", nlevels(factor(training$Deck)), "decks on the ship.")
[1] "Titanic has 8 decks on the ship."
```

```
## $ Family : Factor w/ 875 levels "Abbing - ", "Abbott - ", ...: 101 183 33
5 273 16 544 506 614 388 565 ... set.seed(6)
```

```
> imp = mice(training, method = "rf", m=5)
```

```
iter imp variable
```

```
1 1 Age
1 2 Age
1 3 Age
1 4 Age
1 5 Age
2 1 Age
2 2 Age
2 3 Age
2 4 Age
2 5 Age
3 1 Age
3 2 Age
3 3 Age
3 4 Age
3 5 Age
4 1 Age
4 2 Age
4 3 Age
4 4 Age
4 5 Age
5 1 Age
5 2 Age
5 3 Age
5 4 Age
5 5 Age
```

```
Warning message:
```

```
Number of logged events: 8
```

```
imputedtraining = complete(imp)
```

```
> summary(imp)
```

```
Class: mids
```

```
Number of multiple imputations: 5
```

```
Imputation methods:
```

PassengerId	Survived	Pclass	Name	Sex	Age
SibSp	""	""	""	""	"rf"
""					
Parch	Ticket	Fare	Cabin	Embarked	Salutation
urname	""	""	""	""	""
""					
Deck					

```

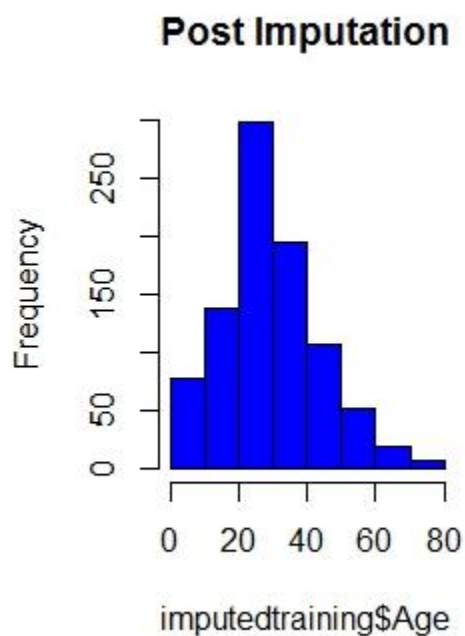
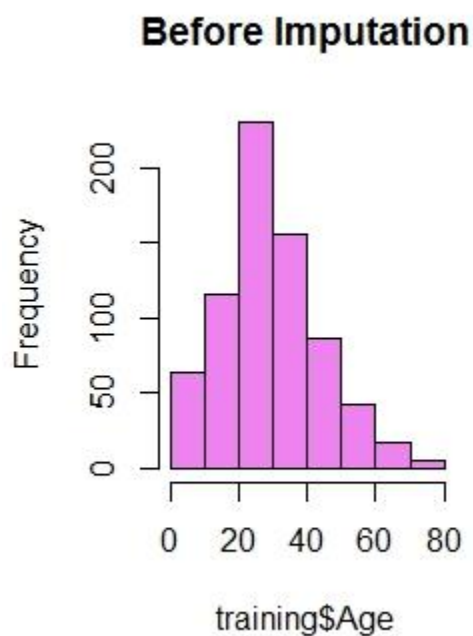
"""
PredictorMatrix:
PassengerId Survived Pclass Name Sex Age SibSp Parch Ticket Fare
Cabin Embarked
PassengerId      0      1      1      0      0      1      1      1      0      1
0      0
Survived          1      0      1      0      0      1      1      1      0      1
0      0
Pclass            1      1      0      0      0      1      1      1      0      1
0      0
Name              1      1      1      0      0      1      1      1      0      1
0      0
Sex                1      1      1      0      0      1      1      1      0      1
0      0
Age                1      1      1      0      0      0      1      1      0      1
0      0
Salutation Surname Deck
PassengerId      0      0      0
Survived          0      0      0
Pclass            0      0      0
Name              0      0      0
Sex                0      0      0
Age                0      0      0
Number of logged events: 8
  it im dep      meth      out
1  0  0      constant      Name
2  0  0      constant      Sex
3  0  0      constant      Ticket
4  0  0      constant      Cabin
5  0  0      constant      Embarked
6  0  0      constant      Salutation
apply(apply(imputedtraining,2,is.na),2,sum)
PassengerId      Survived      Pclass      Name      Sex      Age
SibSp
0      0      0      0      0      0
0
Parch      Ticket      Fare      Cabin      Embarked      Salutation      S
urname
0      0      0      0      687      2      0
0
Deck
687

```

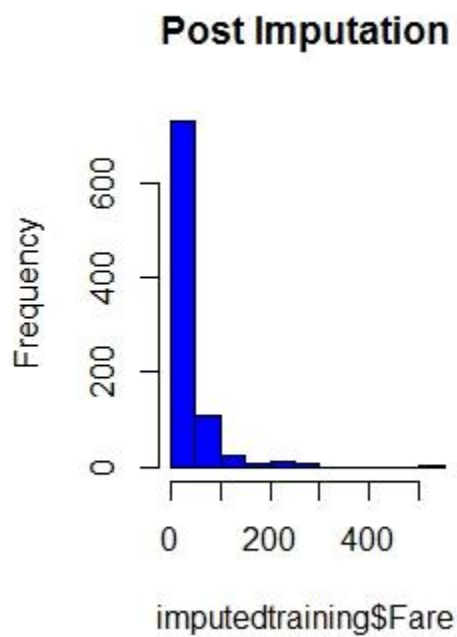
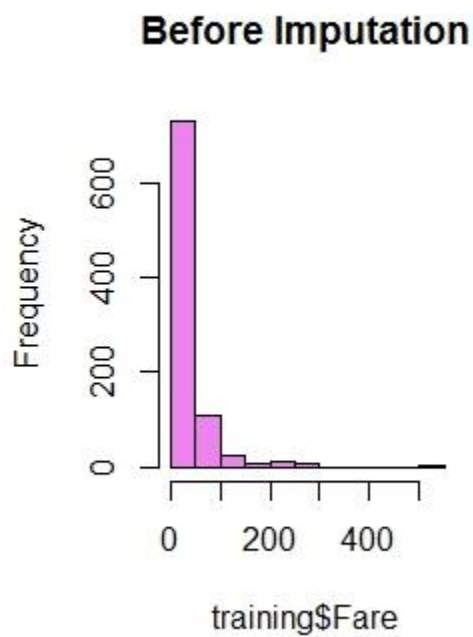
```

par(mfrow=c(1,2))
>
> hist(training$Age, main = "Before Imputation", col = "violet")
> hist(imputedtraining$Age, main = "Post Imputation", col = "blue")

```



```
par(mfrow=c(1,2))
> hist(training$Fare, main = "Before Imputation", col = "violet")
> hist(imputedtraining$Fare, main = "Post Imputation", col = "blue")
```



Missing Values Analysis using Amelia ordered by % missing



#Missing cases (numbers):								
map_int(train.raw,~sum(is.na(.x)))								
##	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket
##	0	0	0	0	117	0	0	0
##	Fare	Cabin	Embarked					
##	0	478	2					

Cabin has a large number of missing values (77% missing). Imputing this variable may prove challenging or even useless. Age (19.9% missing) and Embarked (0.2%) missi

## Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <http://rmarkdown.rstudio.com>.

When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
{r cars} summary(cars)
```

## Including Plots

You can also embed plots, for example:

```
{r pressure, echo=FALSE} plot(pressure)
```

Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.