

PES UNIVERSITY

(Established under Karnataka Act No. 16 of 2013) 100 Ft. Road, BSK III Stage, Bengaluru – 560 085

DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING

Course Title: Image Processing and Data Visualization Using

MATLAB

Course code: -UE19CS257B

SRN: PES1UG19CS321 | Name: PALAK KOTHARI

PROJECT REPORT

Problem Statement: Who Survives the titanic?

Objectives: To build a predictive model that answers the question: "What sorts of people were more likely to survive?" Using passenger data (i.e name, age, gender, socio-economic class, etc)

Description:

About the Dataset:

- Each row in the dataset represents a passenger or a crew member
- Each column has various attributes of the passenger such as passenger's name, age, sex, number of children/parents, number of siblings/spouses, port of embarkation, ticket number, fare, etc.
- There are 891 rows and 12 columns

Description of Dataset:

PASSENGER ID Unique ID used to identify the passenger						
SURVIVED	Value = 1 if passenger survived and Value = 0 if passenger did not survive					

PCLASS	Ticket Class (1=1st, 2=2nd, 3=3rd)						
NAME	Name of the passenger						
SEX	Sex of the passenger						
AGE	Age of the passenger						
SIBSP	Number of siblings/ spouses of the passenger present on the Titanic						
PARCH	Number of parents/children present on the Titanic						
TICKET	Ticket number of the passenger						
FARE	Passenger fare						
CABIN	Cabin number						
EMBARKED	Port of embarkation (C=Cherbourg, Q=Queenstown, S=Southampton)						

Uncleaned Data:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	s
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583	NaN	Q
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625	E46	S
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750	NaN	S
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0	0	2	347742	11.1333	NaN	S
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0	1	0	237736	30.0708	NaN	С

Cleaned Data:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked
0	1	0	3	Braund, Mr. Owen Harris	0	22.0	1	0	A/5 21171	7.2500	0
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	1	38.0	1	0	PC 17599	71.2833	1
2	3	1	3	Heikkinen, Miss. Laina	1	26.0	0	0	STON/O2. 3101282	7.9250	0
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	35.0	1	0	113803	53.1000	0
4	5	0	3	Allen, Mr. William Henry	0	35.0	0	0	373450	8.0500	0

New Concept Learnt (Explanation):

- Data cleaning: Checking for missing Values and Filling them with appropriate average values, and checking for duplicate values and deleting them.
- Data Visualization: Using graphs, we plot various the variables (columns) against the survival rates to try and understand the relationship between each factor.
- Outlier Filtering: How to filter the unwanted outliers, if any.
- Normalization: We normalized the data by calculating the measure of central tendencies for each variable and plotted the new data.
- Hypothesis Testing: We assumed a few hypotheses and ran tests on them to check their applicability.

Learning outcome:

We learnt how to MATLAB's vast functionalities and application in a statistical background.

Handling data and analyzing its various characteristics have never been easier before.

```
disp(countofmissingvalues);
Embarked_mode=mode(Embarked);
ds=fillmissing(ds,'constant',Embarked_mode,"DataVariables",'Embarked');
ix = ismissing(ds);
countofmissingvalues=sum(ix);
disp(countofmissingvalues);
DATA IMPORTING
ds=readtable("titanic_train_uncleaned.csv");
variableNames = ds.Properties.VariableNames;
%ds.(variableNames{9}) = char(ds.(variableNames{9}));
ds.(variableNames{12}) = char(ds.(variableNames{12}));
Survived=ds.Survived:
Pclass=ds.Pclass:
Ticket=ds.Ticket;
Age=ds.Age:
Embarked=ds.Embarked;
%DATA CLEANING
ix = ismissing(ds):
countofmissingvalues=sum(ix);
disp(countofmissingvalues);
ds=fillmissing(ds,'next',"DataVariables",'Ticket');
%ds=fillmissing(ds,'constant',0,"DataVariables",'Ticket');
ix = ismissing(ds);
countofmissingvalues=sum(ix);
disp(countofmissingvalues);
ds=fillmissing(ds,'constant',0,"DataVariables",'Age'); %replace missing values in age with 0 to calculate mean
mean_age=mean(int8(Age));
ds.Age(ds.Age==0)=mean_age; %replace previously placed 0 to mean age value
ix = ismissing(ds);
countofmissingvalues=sum(ix);
disp(countofmissingvalues);
Embarked mode=mode(Embarked):
ds=fillmissing(ds,'constant',Embarked_mode,"DataVariables",'Embarked');
ix = ismissing(ds);
countofmissingvalues=sum(ix);
```

We learnt how to clean, filter and visualize large Datasets that will provide a valuable input to various other fields of Computer Science such as Machine Learning.

We learnt the concept of Hypothesis Testing better and how to apply it practically.

Code:

```
%DATA VISUALISATION
figure(1);
cSurv=categorical(ds.Survived,[0,1],{'0','1'});
histogram(cSurv);
xlabel('Survived');ylabel('Count');
figure(2);
cSex=categorical(ds.Sex,[0,1],{'female','male'});
histogram(cSex, 'FaceColor', 'green');
xlabel('Sex');ylabel('Count');
figure(3);
heatmap(ds,'Sex','Survived');
figure(4);
histogram(ds.Pclass, 'FaceColor', 'red');
xlabel('Pclass');ylabel('Count');
figure(5);
heatmap(ds, 'Survived', 'Pclass');
figure(6);
histogram(ds.Fare, 'FaceColor', 'yellow');xlabel('Fare');
figure(7);
histogram(ds.Age, 'FaceColor', 'cyan');xlabel('Age');
figure(8);
histogram(ds.SibSp,'FaceColor','magenta');xlabel('SibSp');
figure(9);
heatmap(ds,'Survived','SibSp');
figure(10);
histogram(ds.Parch, 'FaceColor', 'None');xlabel('Parch');
figure(11);
heatmap(ds, 'Survived', 'Parch');
```

```
% Fill outliers
   [newTable,outlierIndices,thresholdLow,thresholdHigh] = filloutliers(dscopy,...
       "spline", "movmean", 20, "DataVariables", "Age");
f % Display results
  clf
E
  plot(dscopy.Age, "Color", [77 190 238]/255, "DisplayName", "Input data")
C hold on
  plot(newTable.Age,"Color",[0 114 189]/255,"LineWidth",1.5,...
h
       "DisplayName", "Cleaned data")
f
  % Plot outliers
h
   plot(find(outlierIndices(:,6)),dscopy.Age(outlierIndices(:,6)),"x",...
       "Color",[64 64 64]/255,"DisplayName","Outliers")
d title("Number of outliers: " + nnz(outlierIndices(:,6)))
  % Plot filled outliers
   plot(find(outlierIndices(:,6)),newTable.Age(outlierIndices(:,6)),".",...
       "MarkerSize",12,"Color",[217 83 25]/255,"DisplayName","Filled outliers")
  % Plot outlier thresholds
   plot([(1:numel(dscopy.Age))'; missing; (1:numel(dscopy.Age))'],...
       [thresholdHigh.Age(:); missing; thresholdLow.Age(:)],...
       "Color", [145 145 145]/255, "DisplayName", "Outlier thresholds")
  hold off
   legend
  ylabel("Age")
   clear outlierIndices thresholdLow thresholdHigh
```

```
% Fill outliers
[newTable2,outlierIndices2,thresholdLow2,thresholdHigh2] = filloutliers(newTable,...
    "spline", "movmean", 20, "DataVariables", "Fare");
% Display results
clf
plot(newTable.Fare, "Color", [77 190 238]/255, "DisplayName", "Input data")
hold on
plot(newTable2.Fare, "Color", [0 114 189]/255, "LineWidth", 1.5,...
    "DisplayName", "Cleaned data")
% Plot outliers
plot(find(outlierIndices2(:,10)),newTable.Fare(outlierIndices2(:,10)),"x",...
    "Color",[64 64 64]/255,"DisplayName","Outliers")
title("Number of outliers: " + nnz(outlierIndices2(:,10)))
% Plot filled outliers
plot(find(outlierIndices2(:,10)),newTable2.Fare(outlierIndices2(:,10)),".",...
    "MarkerSize",12,"Color",[217 83 25]/255,"DisplayName","Filled outliers")
% Plot outlier thresholds
plot([(1:numel(newTable.Fare))'; missing; (1:numel(newTable.Fare))'],...
    [thresholdHigh2.Fare(:); missing; thresholdLow2.Fare(:)],...
    "Color", [145 145 145]/255, "DisplayName", "Outlier thresholds")
hold off
legend
ylabel("Fare")
clear outlierIndices2 thresholdLow2 thresholdHigh2
```



```
% Fill outliers
[newTable3,outlierIndices3,thresholdLow3,thresholdHigh3] = filloutliers(newTable2,...
    "spline", "movmean", 20, "DataVariables", "SibSp");
% Display results
plot(newTable2.SibSp,"Color",[77 190 238]/255,"DisplayName","Input data")
hold on
plot(newTable3.SibSp,"Color",[0 114 189]/255,"LineWidth",1.5,...
    "DisplayName", "Cleaned data")
% Plot outliers
plot(find(outlierIndices3(:,7)),newTable2.SibSp(outlierIndices3(:,7)),"x",...
"Color",[64 64 64]/255,"DisplayName","Outliers")
title("Number of outliers: " + nnz(outlierIndices3(:,7)))
% Plot filled outliers
plot(find(outlierIndices3(:,7)),newTable3.SibSp(outlierIndices3(:,7)),".",...
    "MarkerSize",12,"Color",[217 83 25]/255,"DisplayName","Filled outliers")
% Plot outlier thresholds
plot([(1:numel(newTable2.SibSp))'; missing; (1:numel(newTable2.SibSp))'],...
    [thresholdHigh3.SibSp(:); missing; thresholdLow3.SibSp(:)],...
    "Color",[145 145 145]/255,"DisplayName","Outlier thresholds")
hold off
legend
ylabel("SibSp")
clear outlierIndices3 thresholdLow3 thresholdHigh3
```

```
% Fill outliers
[newTable4,outlierIndices4,thresholdLow4,thresholdHigh4] = filloutliers(newTable3,...
    "spline", "movmean", 20, "DataVariables", "Parch");
% Display results
clf
plot(newTable3.Parch, "Color", [77 190 238]/255, "DisplayName", "Input data")
hold on
plot(newTable4.Parch, "Color", [0 114 189]/255, "LineWidth", 1.5,...
    "DisplayName", "Cleaned data")
% Plot outliers
plot(find(outlierIndices4(:,8)),newTable3.Parch(outlierIndices4(:,8)),"x",...
    "Color",[64 64 64]/255,"DisplayName","Outliers")
title("Number of outliers: " + nnz(outlierIndices4(:,8)))
% Plot filled outliers
plot(find(outlierIndices4(:,8)),newTable4.Parch(outlierIndices4(:,8)),".",...
    "MarkerSize",12,"Color",[217 83 25]/255,"DisplayName","Filled outliers")
% Plot outlier thresholds
plot([(1:numel(newTable3.Parch))'; missing; (1:numel(newTable3.Parch))'],...
    [thresholdHigh4.Parch(:); missing; thresholdLow4.Parch(:)],...
    "Color", [145 145 145]/255, "DisplayName", "Outlier thresholds")
hold off
legend
ylabel("Parch")
clear outlierIndices4 thresholdLow4 thresholdHigh4
```

```
%NORMALISATION
disp("Mean of all the columns of the dataset");
d=["Mean of Survived:",mean(nds.Survived)];
disp(d);
d=["Mean of Pclass:",mean(nds.Pclass)];
disp(d);
d=["Mean of Sex:",mean(nds.Sex)];
disp(d);
d=["Mean of Age:",mean(nds.Age)];
disp(d);
d=["Mean of SibSp:",mean(nds.SibSp)];
disp(d);
d=["Mean of Parch:",mean(nds.Parch)];
disp(d);
d=["Mean of Fare:",mean(nds.Fare)];
disp(d);
disp("variance of all the columns of the dataset");
d=["Variance of Survived:",var(nds.Survived)];
disp(d);
d=["Variance of Pclass:",var(nds.Pclass)];
disp(d);
d=["Variance of Sex:",var(nds.Sex)];
disp(d);
d=["Variance of Age:",var(nds.Age)];
disp(d);
d=["Variance of SibSp:",var(nds.SibSp)];
disp(d);
d=["Variance of Parch:",var(nds.Parch)];
disp(d);
d=["Variance of Fare:",var(nds.Fare)];
disp(d);
```

```
data=nds(:,[1:6 8:9 11 7 10]); %rearrange to normalise
data{:, 10:end} = normc(data{:, 10:end}); %fare and age ads the last two cols
%plot after normalisation
figure(14);
histfit(data.Fare);xlabel('Fare');
figure(15);
histfit(data.Age);xlabel('Age');
nds=data;
disp("After normalisation:")
disp("Mean of all the columns of the dataset");
d=["Mean of Survived:",mean(nds.Survived)];
disp(d);
d=["Mean of Pclass:",mean(nds.Pclass)];
disp(d);
d=["Mean of Sex:",mean(nds.Sex)];
disp(d);
d=["Mean of Age:", mean(nds.Age)];
disp(d);
d=["Mean of SibSp:",mean(nds.SibSp)];
disp(d);
d=["Mean of Parch:", mean(nds.Parch)];
disp(d);
d=["Mean of Fare:",mean(nds.Fare)];
disp(d);
disp("variance of all the columns of the dataset");
d=["Variance of Survived:", var(nds.Survived)];
disp(d);
d=["Variance of Pclass:",var(nds.Pclass)];
disp(d):
d=["Variance of Sex:",var(nds.Sex)];
disp(d);
d=["Variance of Age:",var(nds.Age)];
disp(d);
d=["Variance of SibSp:",var(nds.SibSp)];
```

disp(d):

disp(d);

disp(d);

d=["Variance of Parch:",var(nds.Parch)];

d=["Variance of Fare:",var(nds.Fare)];

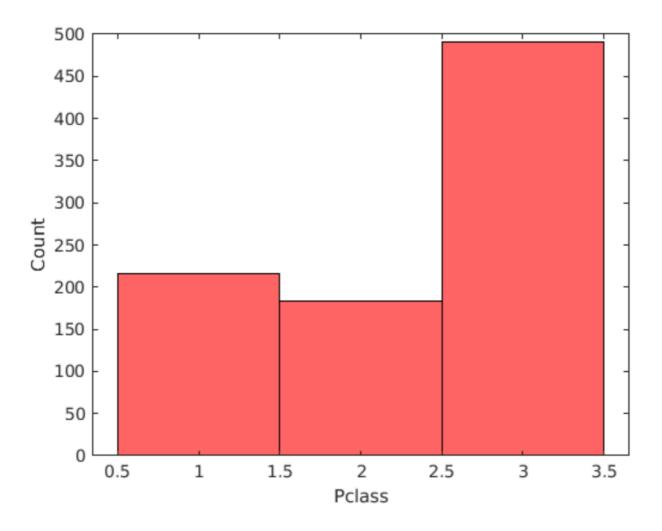
Output Screenshots:

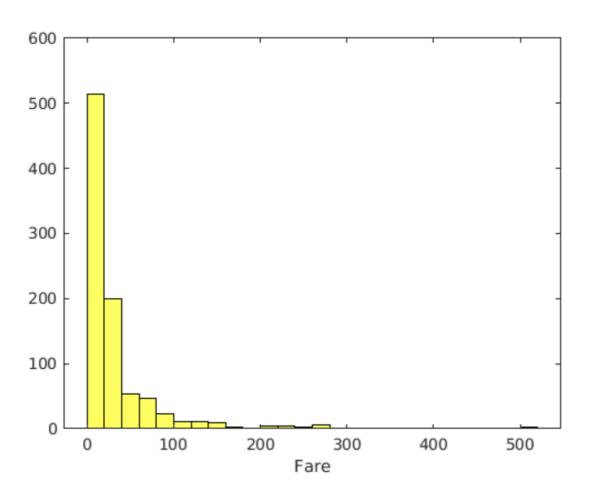
%HYPOTHESIS TESTING

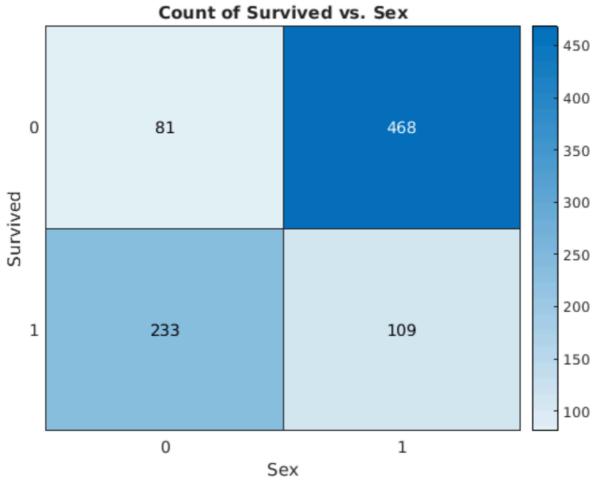
Missing Values and Duplicate Values Handling;

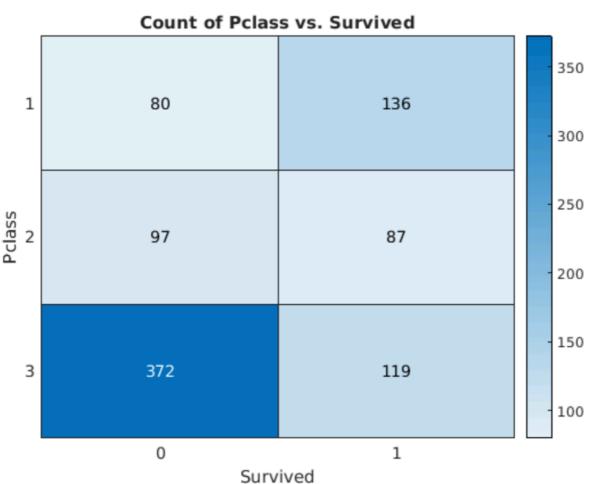
0	0	0	0	0	177	0	0	230	0	687	2
0	0	0	0	0	177	0	0	0	0	687	2
0	0	0	0	0	0	0	0	0	0	687	2
0	0	0	0	0	0	0	0	0	0	687	0
0	0	0	0	0	0	0	0	0	0	0	

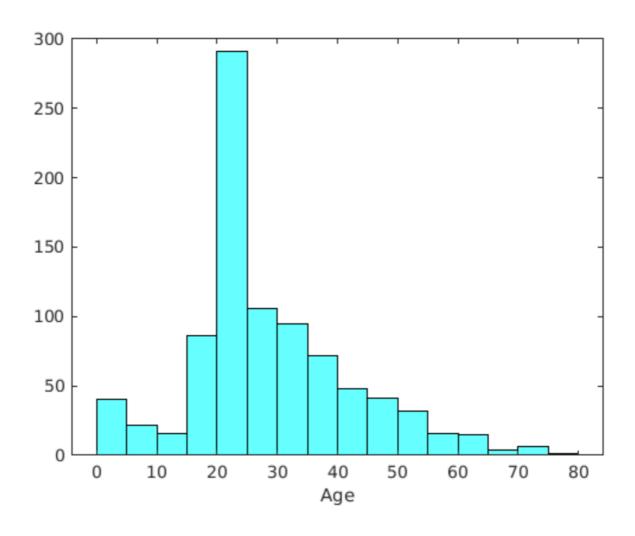
NO DUPLICATES FOUND

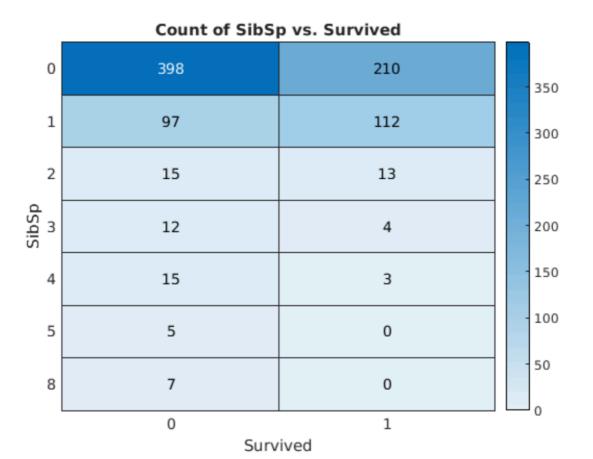


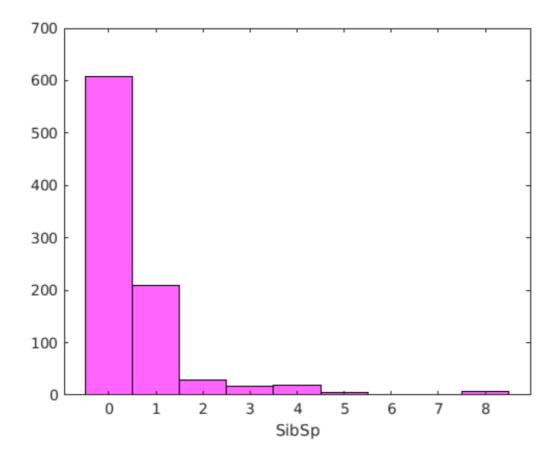


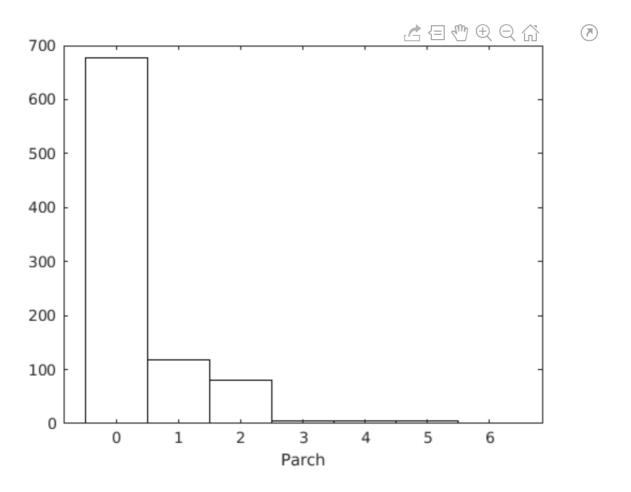


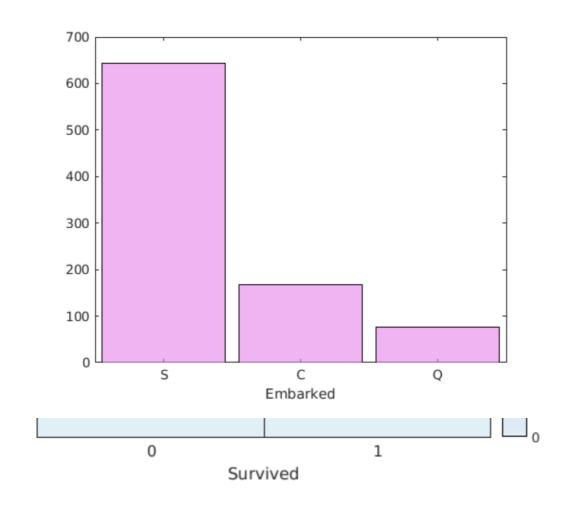


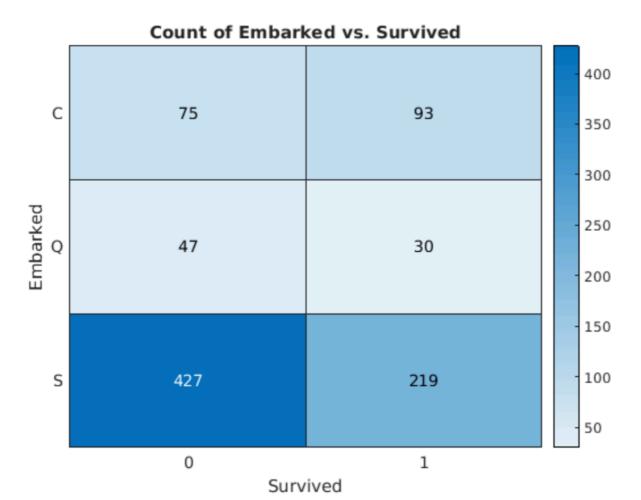


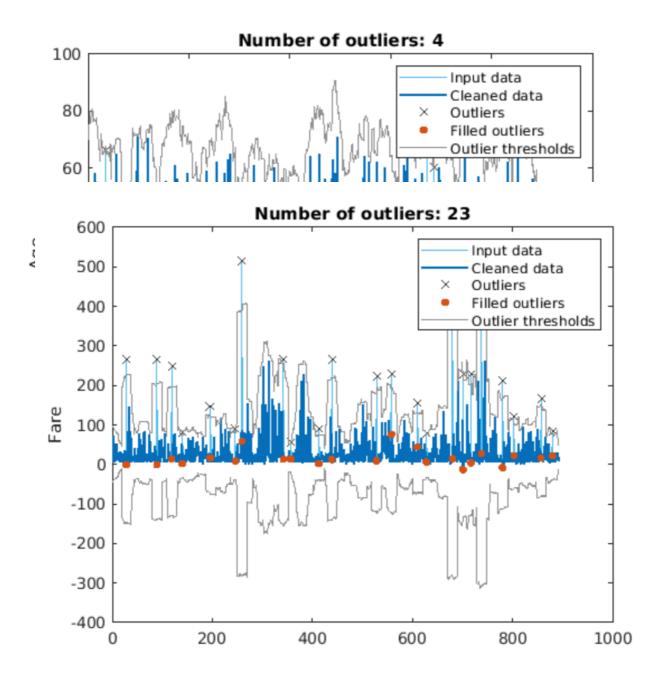


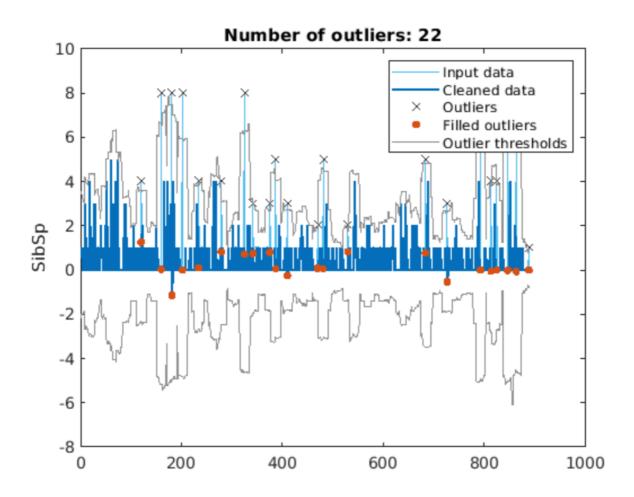


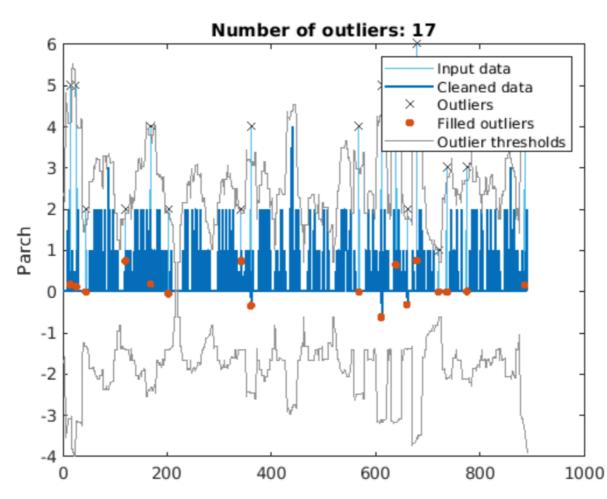












Mean of all the columns of the dataset

"Mean of Survived:" "0.38384"

"Mean of Pclass:" "2.3086"

"Mean of Sex:" "0.64759"

"Mean of Age:" "28.423"

"Mean of SibSp:" "0.40657"

"Mean of Parch:" "0.31679"

"Mean of Fare:" "26.9845"

variance of all the columns of the dataset

"Variance of Survived:" "0.23677"

"Variance of Pclass:" "0.69902"

"Variance of Sex:" "0.22847"

"Variance of Age:" "169.1763"

"Variance of SibSp:" "0.60417"

"Variance of Parch:" "0.41955"

"Variance of Fare:" "1110.3932"

After normalisation:

Mean of all the columns of the dataset
"Mean of Survived:" "0.38384"

"Mean of Pclass:" "2.3086"

"Mean of Sex:" "0.64759"

"Mean of Age:" "0.030466"

"Mean of SibSp:" "0.40657"

"Mean of Parch:" "0.31679"

"Mean of Fare:" "0.02109"

variance of all the columns of the dataset "Variance of Survived:" "0.23677"

"Variance of Pclass:" "0.69902"

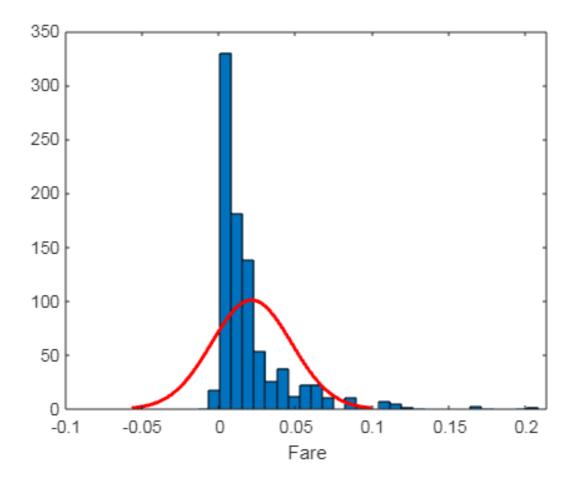
"Variance of Sex:" "0.22847"

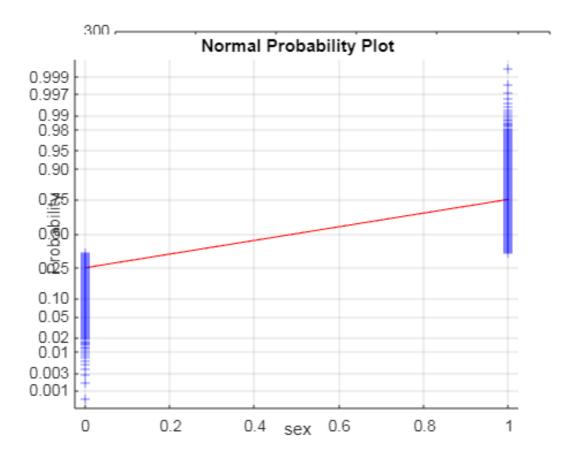
"Variance of Age:" "0.00019437"

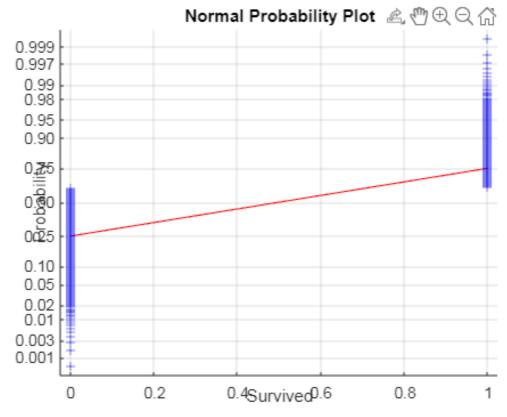
"Variance of SibSp:" "0.60417"

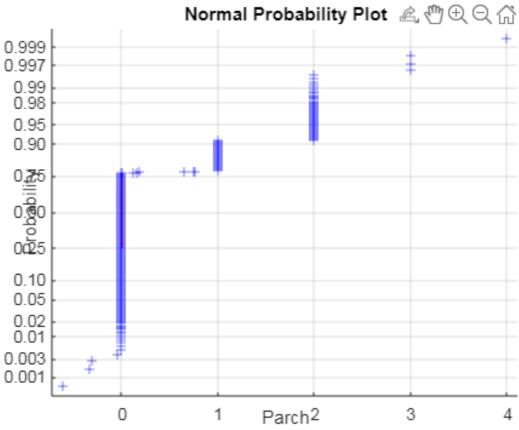
"Variance of Parch:" "0.41955"

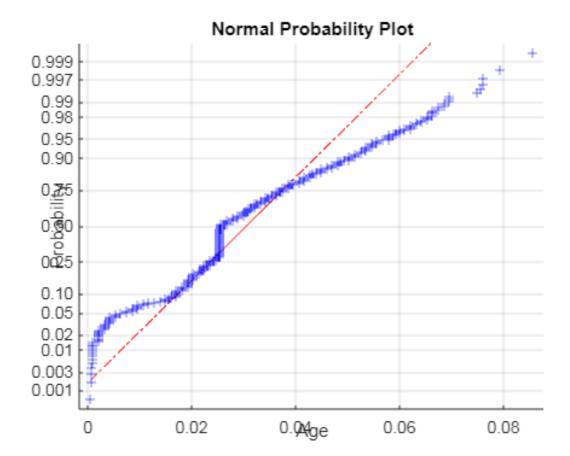
"Variance of Fare:" "0.00067829"

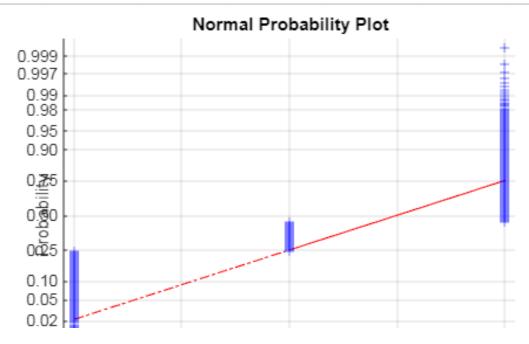












Running a Two-sample Kolmogorov-Smirnov Hypothesis Test; with alpha value 5%

HYPOTHESIS: The proportion of females onboard who survived the sinking of the Titanic was higher than the proportion of males onboard who survived the sinking of the Titanic.

NULL HYPOTHESIS: There is no relationship between the sex and the survived or The proportion of female survivors is equal to the proportion of male survivors

ALTERNATE HYPOTHESIS: There is a relationship between the sex and the survived or The proportion of female survivors is not equal to the proportion of male survivors

The null hypothesis is rejected!!! The sex and survived class are related!

Running a Two-sample Kolmogorov-Smirnov Hypothesis Test: with alpha value 5% HYPOTHESIS: The proportion of females onboard who survived the sinking of the Titanic was higher than the proportion of males onboard who survived the sinking of the Titanic.

NULL HYPOTHESIS: There is no relationship between the sex and the survived or The proportion of female survivors is equal to the proportion of male survivors ALTERNATE HYPOTHESIS: There is a relationship between the sex and the survived or The proportion of female survivors is not equal to the proportion of male survivors

The null hypothesis is rejected!!! The sex and survived class are related!

Correlation Matrix Survi -0.04 -0.54 -0.34 0.03 0.14 0.27 . . . ----0 ----. . . . 4 -0.540.13 -0.12 -0.22 0.07 -0.23 2 & 0 ... Pclas 86426420 -0.340.13 -0.36 0.01 -0.03 -0.57 SibSp 0,03 -0.12 0.21 **9**.18 0.01 Parch 0 7 4 .0.26 .-0.22 °-0.03 0.14 0.370.2680 -0.36 -0.26 0.07 0.1 -0.040.21 0.1 0.05 0.2 0.26 0,11 -0.57 27 -0.23 0.180.1 0 0 0 0.1 0.2 4 6 8 0 2 4 6 2 40 4 0 0.1 0 2 Survi Sex SibSp Parch Pclas Fare Age

Name and signature of the faculty: Prof. Revathi