```
In [2]: import numpy as np import pandas as pd
```

```
In [3]: # Loading data
titanic = pd.read_csv('titanic_data.csv')
```

```
In [5]: def count passenger(df):
            return len(df)
        def remove_column(df, key):
            new_df = df.copy()
            return new_df.drop(key, axis=1)
        def had family(df):
            if df > 0:
                 return 'With family'
            else:
                 return 'Alone'
        def had_alive(df):
            if df > 0:
                 return "alive"
            else:
                 return "died"
        def had class(df):
            if df == 1:
                 return 'First'
            elif df == 2 :
                 return 'Second'
            else:
                 return 'Third'
        def had_child(passenger):
            age, sex = passenger
            if age <= 15:
                 return 'Child'
            else:
                 return sex
```

```
In [8]: # find all the unique values for "Age"
         print(titanic['Age'].unique())
         [ 22.
                  38.
                          26.
                                  35.
                                         28.
                                                 54.
                                                         2.
                                                                27.
                                                                       14.
                                                                                4.
         58.
           20.
                  39.
                          55.
                                  31.
                                         34.
                                                 15.
                                                         8.
                                                                19.
                                                                       40.
                                                                               66.
         42.
           21.
                  18.
                           3.
                                  7.
                                         49.
                                                 29.
                                                        65.
                                                                28.5
                                                                        5.
                                                                               11.
         45.
                                 25.
                                         0.83 30.
                                                                               46.
           17.
                  32.
                          16.
                                                        33.
                                                                23.
                                                                       24.
         59.
           71.
                  37.
                          47.
                                  14.5
                                         70.5
                                                 32.5
                                                                 9.
                                                                       36.5
                                                                               51.
                                                        12.
         55.5
           40.5
                  44.
                          1.
                                  61.
                                         56.
                                                 50.
                                                        36.
                                                                45.5
                                                                       20.5
                                                                               62.
         41.
                          23.5
                               0.92 43.
                                                 60.
                                                        10.
                                                                64.
                                                                               48.
           52.
                  63.
                                                                       13.
         0.75
                          80.
                                  70.
                                         24.5
                                                  6.
                                                         0.67
                                                                30.5
                                                                        0.42
                                                                               34.5
           53.
                  57.
        74. ]
```

In [9]: # fill missing "Age" with mean
 titanic['Age'] = titanic['Age'].fillna(titanic['Age'].median())

In [10]: # find all the unique values for "Embarked"
print titanic['Embarked'].unique()

['S' 'C' 'Q' nan]

In [11]: # replace all the missing values in the Embarked column with S.
titanic['Embarked'] = titanic['Embarked'].fillna('S')

In [12]: # find all the unique values for "Cabin" print titanic['Cabin'].unique()

```
[nan 'C85' 'C123' 'E46' 'G6' 'C103' 'D56' 'A6' 'C23 C25 C27' 'B78' 'D3
3 '
 'B30' 'C52' 'B28' 'C83' 'F33' 'F G73' 'E31' 'A5' 'D10 D12' 'D26' 'C11
 'B58 B60' 'E101' 'F E69' 'D47' 'B86' 'F2' 'C2' 'E33' 'B19' 'A7' 'C49'
 'A32' 'B4' 'B80' 'A31' 'D36' 'D15' 'C93' 'C78' 'D35' 'C87' 'B77' 'E67
 'B94' 'C125' 'C99' 'C118' 'D7' 'A19' 'B49' 'D' 'C22 C26' 'C106' 'C65'
 'E36' 'C54' 'B57 B59 B63 B66' 'C7' 'E34' 'C32' 'B18' 'C124' 'C91' 'E4
 'T' 'C128' 'D37' 'B35' 'E50' 'C82' 'B96 B98' 'E10' 'E44' 'A34' 'C104'
 'C111' 'C92' 'E38' 'D21' 'E12' 'E63' 'A14' 'B37' 'C30' 'D20' 'B79' 'E
25'
 'D46' 'B73' 'C95' 'B38' 'B39' 'B22' 'C86' 'C70' 'A16' 'C101' 'C68' 'A
10'
 'E68' 'B41' 'A20' 'D19' 'D50' 'D9' 'A23' 'B50' 'A26' 'D48' 'E58' 'C12
6 '
 'B71' 'B51 B53 B55' 'D49' 'B5' 'B20' 'F G63' 'C62 C64' 'E24' 'C90' 'C
 'E8' 'B101' 'D45' 'C46' 'D30' 'E121' 'D11' 'E77' 'F38' 'B3' 'D6' 'B82
B84'
 'D17' 'A36' 'B102' 'B69' 'E49' 'C47' 'D28' 'E17' 'A24' 'C50' 'B42' 'C
148'1
```

break

In [17]: titanic.head()

Out[17]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22	1	0	A/5 21171	7.25(
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38	1	0	PC 17599	71.28
2	3	1	3	Heikkinen, Miss. Laina	female	26	0	0	STON/O2. 3101282	7.92
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	1	0	113803	53.10
4	5	0	3	Allen, Mr. William Henry	male	35	0	0	373450	8.050

```
In [30]:
         # add "Passenger" column represent who alive or died
         titanic['Passenger'] = map(had alive, titanic['Survived'])
         # add "Gender" column these map Sex from text to number for future ana
         lysis
         titanic['Gender'] = titanic['Sex'].map({'female' : 0, 'male' : 1}).ast
         ype(int)
         # add "Who" column, to categorize passenger by group of male, female
         # and children as who under 15 as a child,
         titanic['Who'] = titanic[['Age','Sex']].apply(had child, axis=1)
         # add "Port" column these map each embarktation from text to number fo
         r future analysis
         titanic['Port'] = titanic['Embarked'].dropna().map({'C': 0, 'Q': 1,
         'S': 2 }).astype(int)
         # add "Class" column represen clss of each passenger
         titanic['Class'] = map(had class, titanic['Pclass'])
         # add "Family" column to represent passenger who travel alone or with
         their family
         titanic['Family'] = map(had family, titanic['Parch'] + titanic['SibSp'
         1)
         # drop "Parch" & "SibSp"
         titanic = remove column(titanic, ['Parch', 'SibSp'])
```

In [31]: titanic.head()

Out[31]:

	PassengerId	Survived	Pclass	Name	Sex	Age	Ticket	Fare	Cabin	Em
0	1	0	3	Braund, Mr. Owen Harris	male	22	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35	373450	8.0500	NaN	S

```
In [32]: import math
```

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from pylab import rcParams

from pandas.tools.plotting import scatter matrix

sns.set(style="white")

%matplotlib inline

//anaconda/lib/python2.7/site-packages/matplotlib/__init__.py:872: Use rWarning: axes.color_cycle is deprecated and replaced with axes.prop_c ycle; please use the latter.

warnings.warn(self.msg_depr % (key, alt_key))

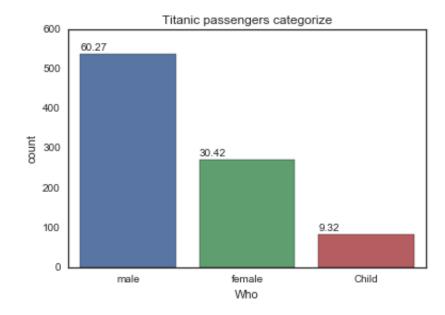
```
In [34]: passengercount = float(len(titanic))

# Investigate who were the passenger

passenger_by_who = titanic.groupby('Who')
print passenger_by_who.count()['Passenger']
sns.plt.title('Titanic passengers categorize')
ax = sns.countplot(x = 'Who', data = titanic)
for p in ax.patches:
    height = p.get_height()
    ax.text(p.get_x(), height + 10, '%1.2f'%((height*100)/passengercount))
```

Who Child 83 female 271 male 537

Name: Passenger, dtype: int64



```
In [35]: # Investigate passengers by class
# show passenger count
passenger_by_who_class = titanic.groupby(['Who','Class'])
print passenger_by_who_class.count()['Passenger']

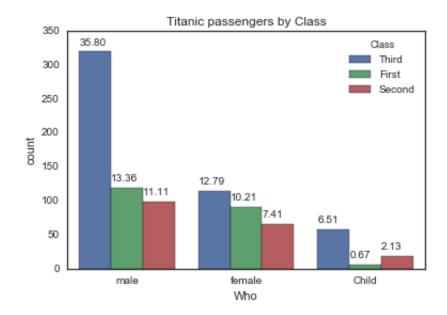
# set plot title
ax = sns.plt.title('Titanic passengers by Class')

# show the counts of passengers
ax = sns.countplot(x = 'Who', hue = 'Class', data = titanic)

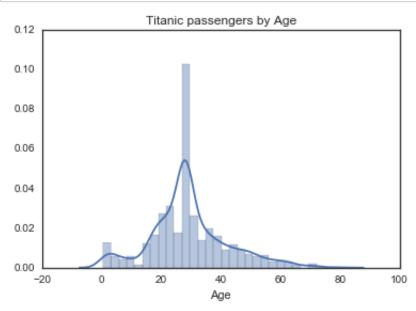
# add percentage for each group
for p in ax.patches:
    height = p.get_height()
    ax.text(p.get_x(), height + 10, '%1.2f'%((height*100)/passengercount))
```

Who Class Child First 6 Second 19 Third 58 female First 91 Second 66 Third 114 First male 119 Second 99 319 Third

Name: Passenger, dtype: int64

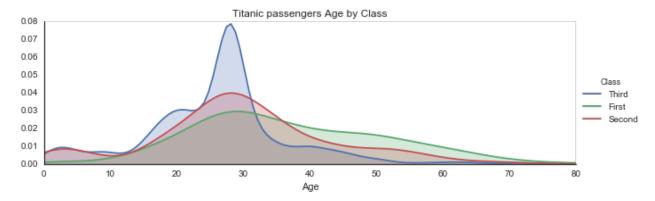


```
In [36]: # quick look the age distribution
    ax = sns.distplot(titanic['Age'], hist = True)
    ax = sns.plt.title('Titanic passengers by Age')
```



```
In [37]: # check age for passengers by class

ax = sns.FacetGrid(titanic, hue = 'Class', aspect = 3)
ax.map(sns.kdeplot,'Age',shade = True)
oldest = titanic['Age'].max()
ax.set(xlim=(0,oldest))
ax.add_legend();
ax = sns.plt.title('Titanic passengers Age by Class')
```

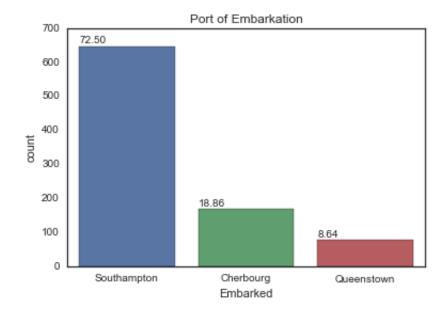


```
In [39]: # Check passengers who from
    titanic['Embarked'] = titanic['Embarked'].dropna().map({'C' : 'Cherbou
    rg', 'Q' : 'Queenstown', 'S': 'Southampton' })
    passenger_by_embark_who = titanic.groupby('Embarked')

print passenger_by_embark_who.count()['Passenger']
    ax = sns.plt.title('Port of Embarkation')
    ax = sns.countplot(x = 'Embarked', data = titanic)
    for p in ax.patches:
        height = p.get_height()
        ax.text(p.get_x(), height + 10, '%1.2f'%((height*100)/passengercou
    nt))
```

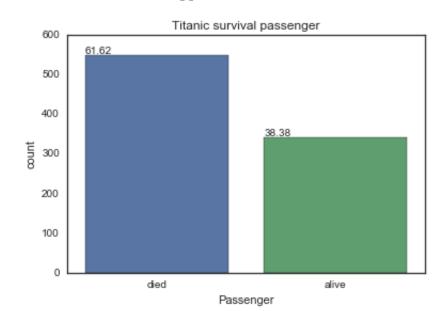
Embarked
Cherbourg 168
Queenstown 77
Southampton 646

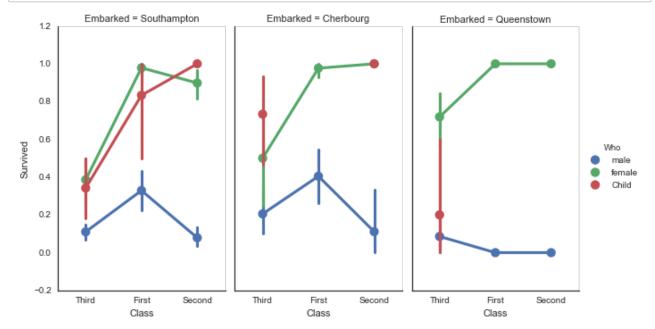
Name: Passenger, dtype: int64

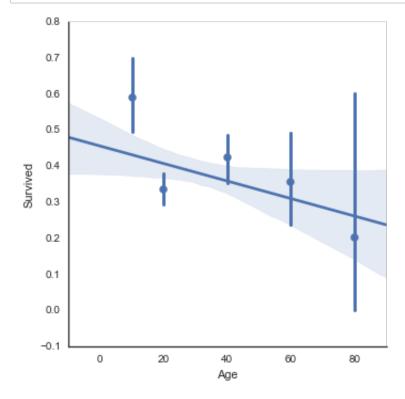


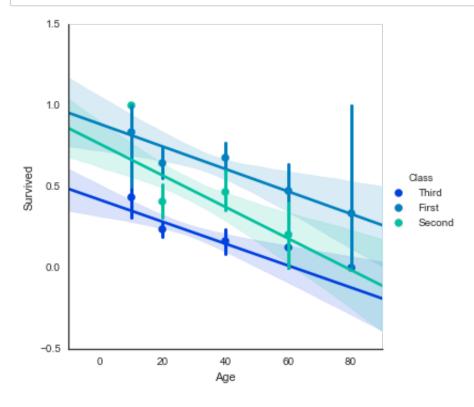
```
In [40]: # check survial rate
    passenger_by_class_who = titanic.groupby('Passenger')
    print passenger_by_class_who.count()['Survived']
    ax = sns.plt.title('Titanic survival passenger')
    ax = sns.countplot(x = 'Passenger', data = titanic)
    for p in ax.patches:
        height = p.get_height()
        ax.text(p.get_x(), height + 2, '%1.2f'%((height*100)/passengercount))
```

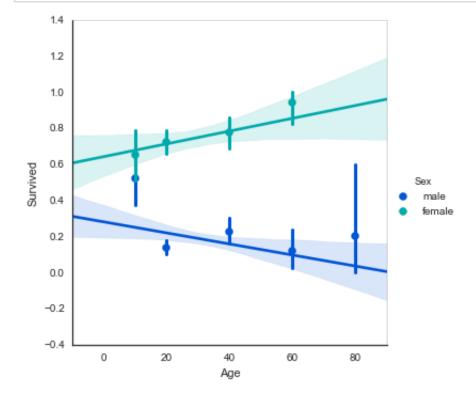
Passenger alive 342 died 549 Name: Survived, dtype: int64

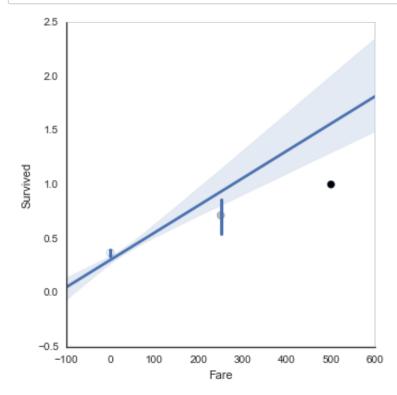


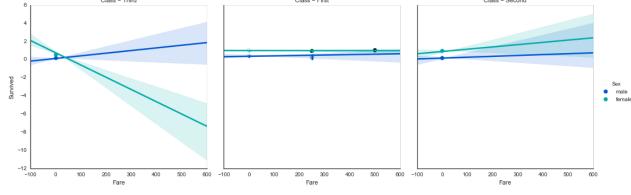


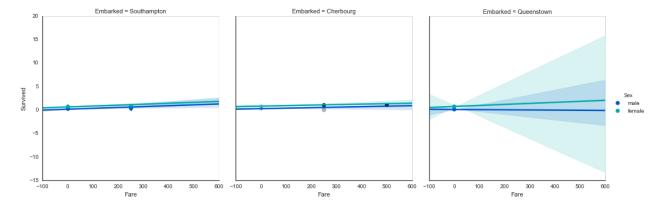












In []: