

```
In [1]: file = open('adult.data', 'r')
```

```
In [2]: def chr_int(a):
        if a.isdigit():
            return int(a)
        else:
            return 0

        data=[]
        for line in file:
            data1=line.split(',')
            if len(data1)==15:
                data.append([chr_int(data1[0]),data1[1],chr_int(data1[2]),data1[3],chr_int(data1[4]),
                            data1[7],data1[8],data1[9],chr_int(data1[10]),chr_int(data1[11]),chr_int(data1[12]),
                            data1[14])])
```

```
In [3]: print data[1:2]

[[50, 'Self-emp-not-inc', 83311, 'Bachelors', 13, 'Married-civ-spouse', 'Exec-managerial', 'Husband', 'White', 'Male', 0, 0, 13, 'United-States', '<=50K\n']]
```

```
In [4]: %matplotlib inline
import pandas as pd

df = pd.DataFrame(data) # Two-dimensional size-mutable, potentially heterogeneous tabular data structure
df.columns = ['age', 'type_employer', 'fnlwgt', 'education',
              'education_num', 'marital', 'occupation', 'relationship', 'race', 'sex',
              'capital_gain', 'capital_loss', 'hr_per_week', 'country', 'income']
df.head()
```

```
Out[4]:
```

	age	type_employer	fnlwgt	education	education_num	marital	occupation	relationship	race	sex
0	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male
1	50	Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male
2	38	Private	215646	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	Male
3	53	Private	234721	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male
4	28	Private	338409	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	Female

```
In [5]: m1 = df[(df.sex == 'Male')] # grouping by sex
m1.shape
```

```
Out[5]: (21790, 15)
```

```
In [6]: m11 = df[(df.sex == 'Male') & (df.income == '>50K\n')]
m11.shape
```

```
Out[6]: (6662, 15)
```

```
In [7]: fm = df[(df.sex == 'Female')]
fm.shape
```

```
Out[7]: (10771, 15)
```

```
In [8]: fm1 =df[(df.sex == 'Female')&(df.income=='>50K\n')]  
fm1.shape
```

```
Out[8]: (1179, 15)
```

Calculating mean, standard deviation and variance using numpy

```
In [9]: import numpy as np  
average_age_men = np.mean(m1['age'])  
print average_age_men
```

```
39.4335474989
```

```
In [10]: std_age_men = np.std(m1['age'])  
print std_age_men
```

```
13.3703233824
```

```
In [11]: var_age_men = np.var(m1['age'])  
print var_age_men
```

```
178.765547351
```

```
In [12]: median_age_men = np.median(m1['age'])  
print median_age_men
```

```
38.0
```

```
In [13]: upper_quartile_age_men = np.percentile(m1['age'],75)  
print upper_quartile_age_men
```

```
48.0
```

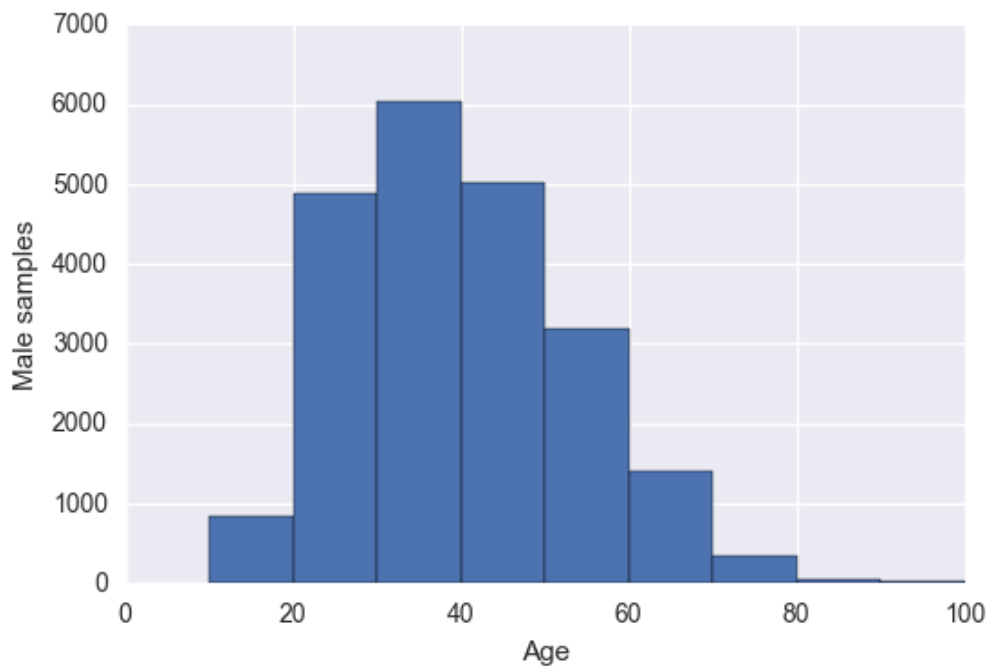
```
In [14]: lower_quartile_age_men = np.percentile(m1['age'],25)  
print lower_quartile_age_men
```

```
29.0
```

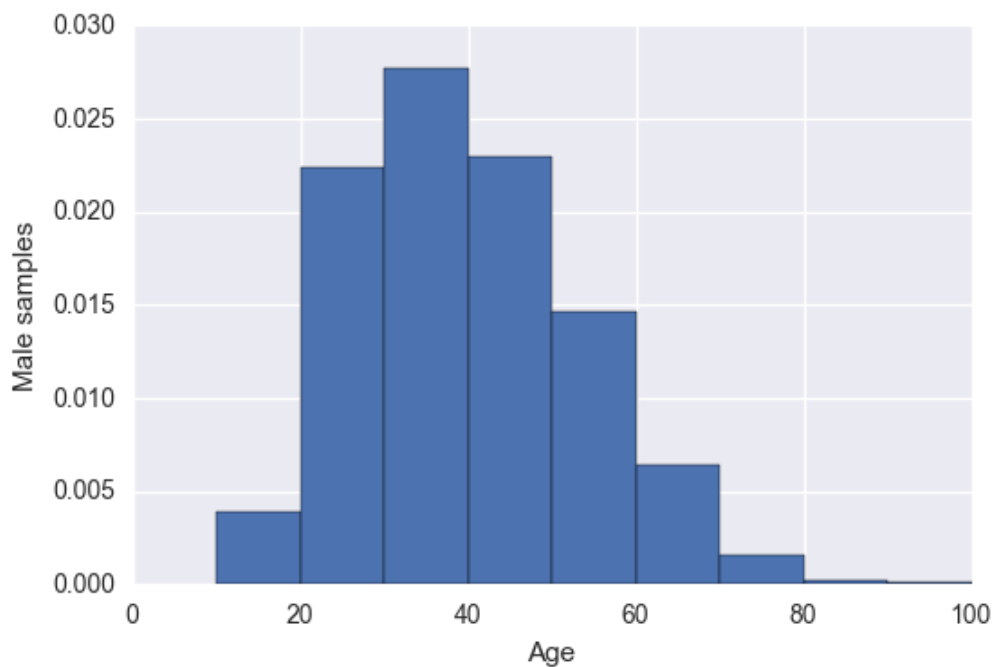
```
In [15]: iqr_age_men = upper_quartile_age_men - lower_quartile_age_men  
print iqr_age_men
```

```
19.0
```

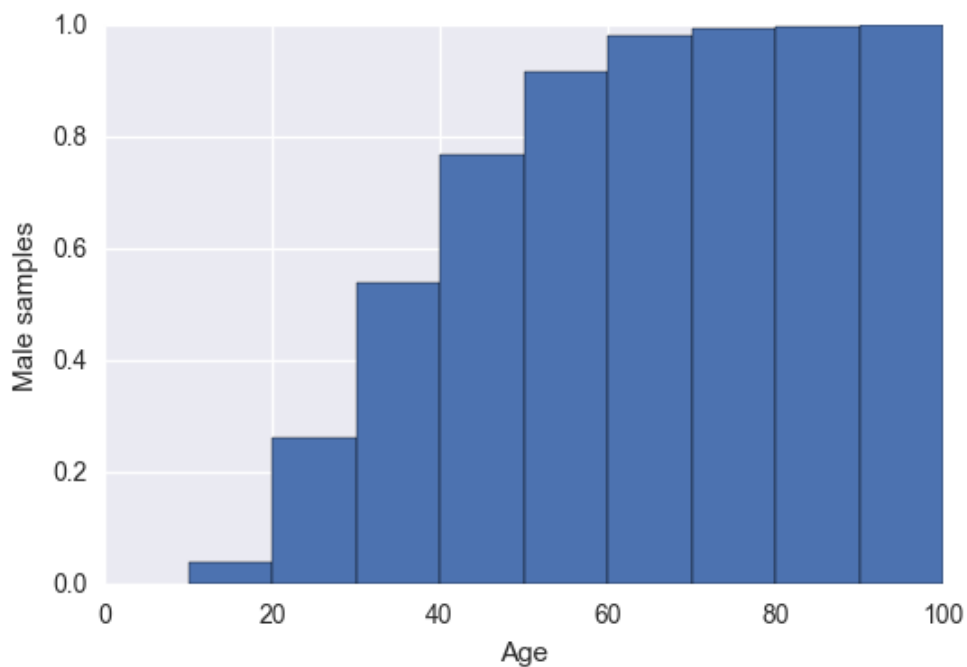
```
In [16]: import matplotlib.pyplot as plt
import seaborn as sns
sns.set()
bin_edges = [0, 10, 20, 30, 40, 50, 60, 70, 80, 90, 100]
_ = plt.hist(ml['age'], bins = bin_edges)
_ = plt.xlabel('Age')
_ = plt.ylabel('Male samples')
plt.show()
```



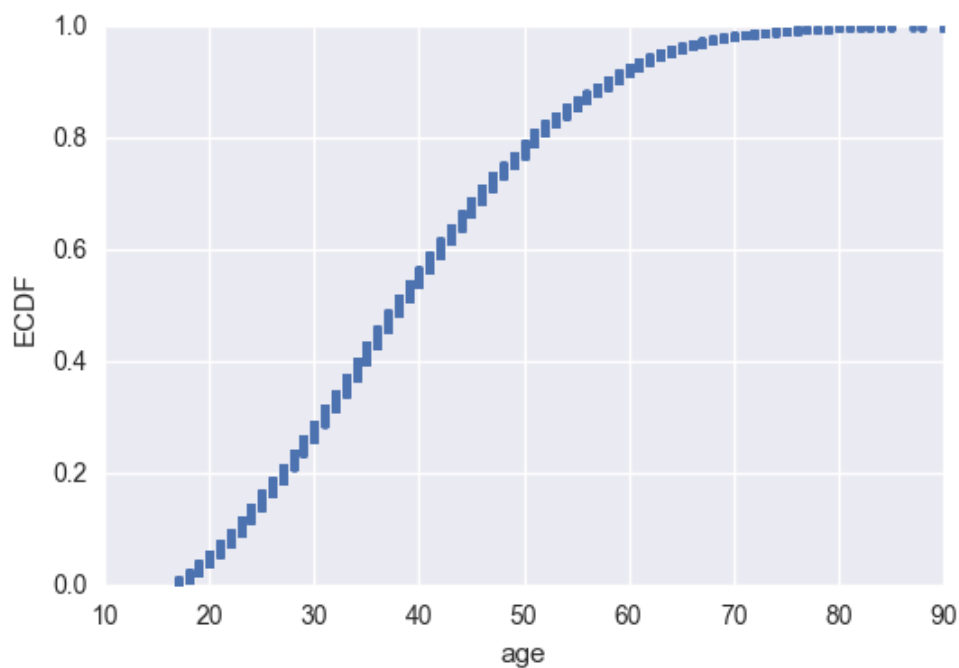
```
In [17]: import matplotlib.pyplot as plt
bin_edges = [0, 10, 20, 30, 40, 50, 60, 70, 80, 90, 100]
_ = plt.hist(ml['age'], bins=bin_edges, normed = 1)
_ = plt.xlabel('Age')
_ = plt.ylabel('Male samples')
plt.show()
```



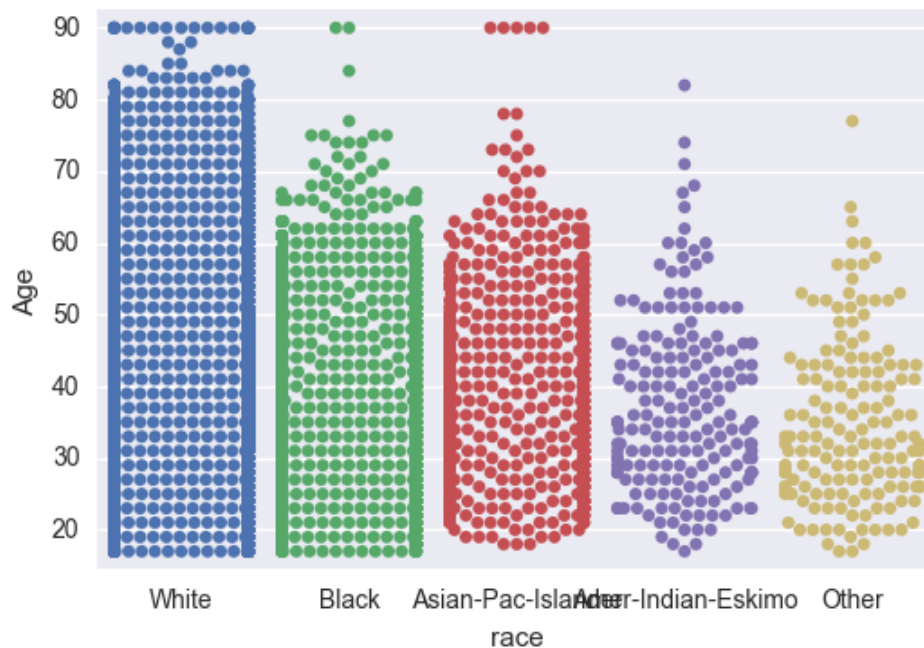
```
In [18]: _ = plt.hist(ml['age'], bins=bin_edges, cumulative=True, normed = 1)
_ = plt.xlabel('Age')
_ = plt.ylabel('Male samples')
plt.show()
```



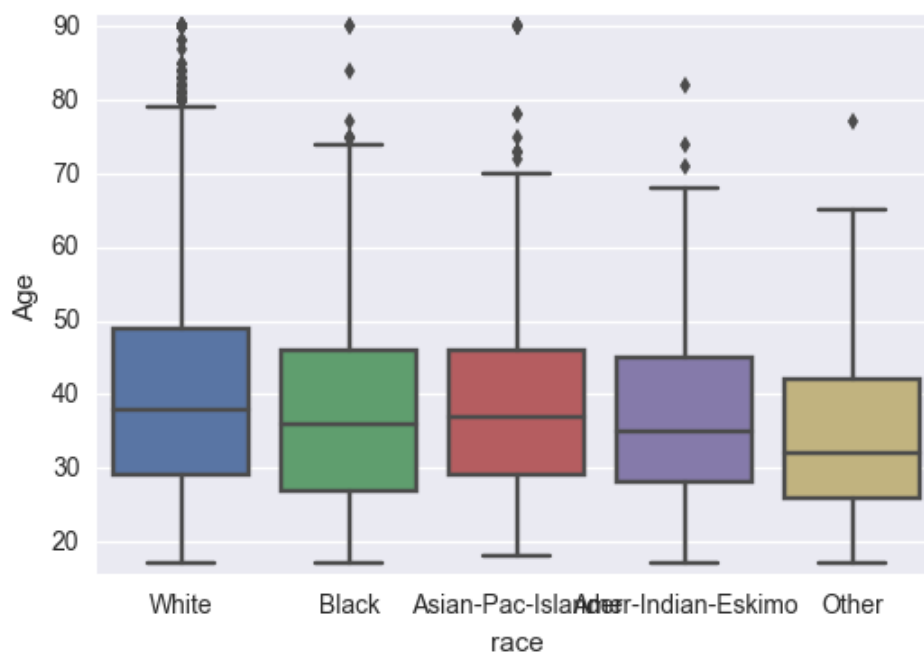
```
In [19]: x = np.sort (ml['age'])
y = np.arange(1.0, len(x)+1)/len(x)
_ = plt.plot(x, y, marker = '.', linestyle = 'none')
_ = plt.xlabel ('age')
_ = plt.ylabel ('ECDF')
plt.show()
```



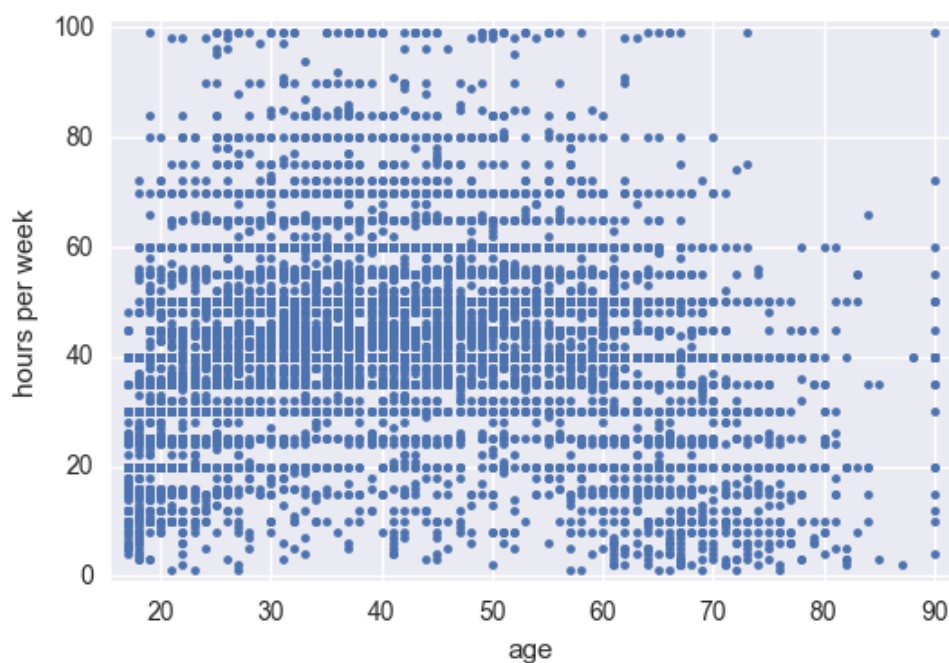
```
In [20]: import seaborn as sns
sns.set()
_ = sns.swarmplot(x= 'race', y = 'age', data = m1)
_ = plt.margins(0.02)
_ = plt.xlabel ('race')
_ = plt.ylabel ('Age')
plt.show()
```



```
In [21]: import seaborn as sns
sns.set()
_ = sns.boxplot(x = 'race', y = 'age', data = m1)
_ = plt.margins(0.02)
_ = plt.xlabel ('race')
_ = plt.ylabel ('Age')
plt.show()
```



```
In [22]: _ = plt.plot(ml['age'], ml['hr_per_week'], marker = '.', linestyle = 'none')
_ = plt.margins(0.02)
_ = plt.xlabel('age')
_ = plt.ylabel('hours per week')
plt.show()
```

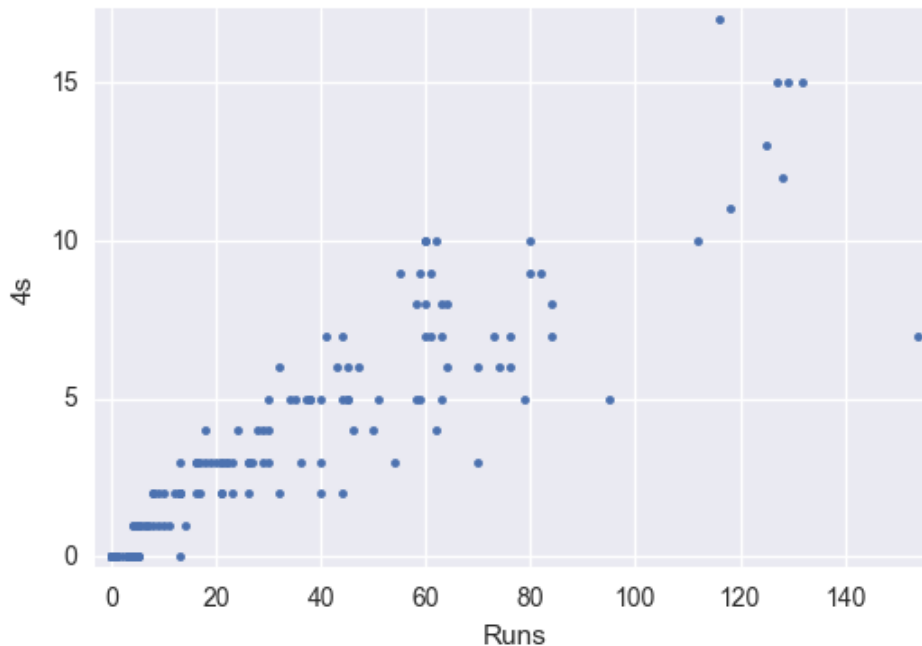


```
In [23]: batsman = pd.read_excel('batsman.xlsx')
batsman.head()
```

```
Out[23]:
```

	Runs	Mins	BF	4s	6s	SR	Pos	Dismissal	Inns	Start Date
1	5	10	8	0	0	62.50	1	caught	1	2007-02-09
2	30	40	32	3	2	93.75	1	caught	2	2007-02-10
3	7	27	20	1	0	35.00	2	run out	2	2007-04-19
4	45	81	53	6	0	84.90	2	caught	1	2007-05-10
5	7	22	12	1	0	58.33	2	run out	2	2007-05-12

```
In [24]: _ = plt.plot(batsman['Runs'], batsman['4s'], marker = '.', linestyle = 'none')
_ = plt.margins(0.02)
_ = plt.xlabel('Runs')
_ = plt.ylabel('4s')
plt.show()
```



```
In [25]: covariance_matrix = np.cov(ml['hr_per_week'],ml['age'])
print covariance_matrix
print "Covariance = ", covariance_matrix [0][1]
```

```
[[ 146.88846717   5.5298026 ]
 [   5.5298026  178.77375175]]
Covariance = 5.52980260364
```

```
In [26]: pearson_coefficient = np.corrcoef(ml['hr_per_week'],ml['age'])
print pearson_coefficient
print "Pearson's correlation coefficient = " , pearson_coefficient [0][1]
```

```
[[ 1.          0.03412431]
 [ 0.03412431  1.          ]]
Pearson's correlation coefficient = 0.0341243118412
```

```
In [27]: covariance_matrix1 = np.cov(batsman['Runs'], batsman['4s'])
print covariance_matrix1
print "Covariance between runs and 4s= ", covariance_matrix1 [0][1]
```

```
[[ 1169.20642119  112.07718121]
 [ 112.07718121  13.16724107]]
Covariance between runs and 4s= 112.077181208
```

```
In [28]: pearson_coefficient1 = np.corrcoef(batsman['Runs'], batsman['4s'])
print pearson_coefficient1
print "Pearson correleation coefficient between runs and 4s= ", pearson_coefficient1 [0][1]
```

```
[[ 1.          0.90328378]
 [ 0.90328378  1.          ]]
Pearson correleation coefficient between runs and 4s= 0.90328377718
```