

Reading the .csv file and creating a dataframe using pandas library

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import scipy
from scipy import stats
```

```
df = pd.read_csv("43.csv")
df.head()
```

	gender	race	parental level of education	lunch	\
0	female	group B	bachelor's degree	standard	
1	female	group C	some high school	free/reduced	
2	male	group C	master's degree	free/reduced	
3	female	group C	high school	free/reduced	
4	female	group C	associate's degree	free/reduced	

	test preparation course	math score	reading score	writing score
0	none	99.0	85.0	88.0
1	none	96.0	103.0	102.0
2	none	117.0	29.0	107.0
3	completed	74.0	70.0	58.0
4	none	103.0	98.0	89.0

Data cleaning

handling impossible values (replacing with nan) --solution to first half of introductory question 2

```
#checking how many values greater than 100
(df["math score"]>100).sum()##----309 VALUES
```

```
#checking how many values less than 0
(df["math score"]<0).sum() ##----0 VALUES
```

```
df['math score']=df['math score'].apply(lambda x: np.nan if (x > 100
or x < 0) else x)
```

```
#do the same for reading score and writing score
```

```
df['reading score']=df['reading score'].apply(lambda x: np.nan if (x >
100 or x < 0) else x)
```

```
df['writing score']=df['writing score'].apply(lambda x: np.nan if (x >
100 or x < 0) else x)
```

replacing Nan values in marks with respective means

```
mathscoremean = df['math score'].mean()
readingscoremean = df['reading score'].mean()
writingscoremean = df['writing score'].mean()

df['math score'].fillna(mathscoremean,inplace=True)
df['reading score'].fillna(readingscoremean, inplace=True)
df['writing score'].fillna(writingscoremean, inplace=True)
df['math score']
```

```
0      99.000000
1      96.000000
2      83.587464
3      74.000000
4      83.587464
...
995     83.587464
996     89.000000
997     86.000000
998     95.000000
999     83.587464
Name: math score, Length: 1000, dtype: float64
```

calculating and adding percentages column

--solution to introductory question 1 -- done by calculating sum of marks columns and dividing by 3

```
df['percentage'] =df.iloc[:, -3:-1].sum(axis=1)/3
```

Removing any row with missing categorical data --solution to second half of introductory question 2

```
df.dropna(subset=['gender', 'race', 'parental level of education',
'lunch',
               'test preparation course'],inplace=True)
```

```
df['reading score'].min()
```

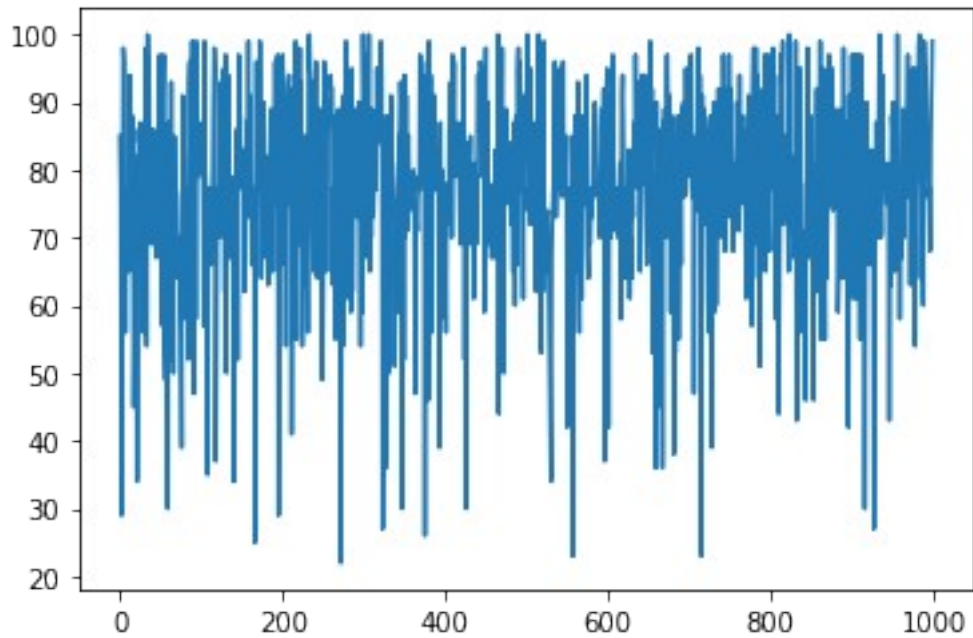
22.0

Plotting the reading score

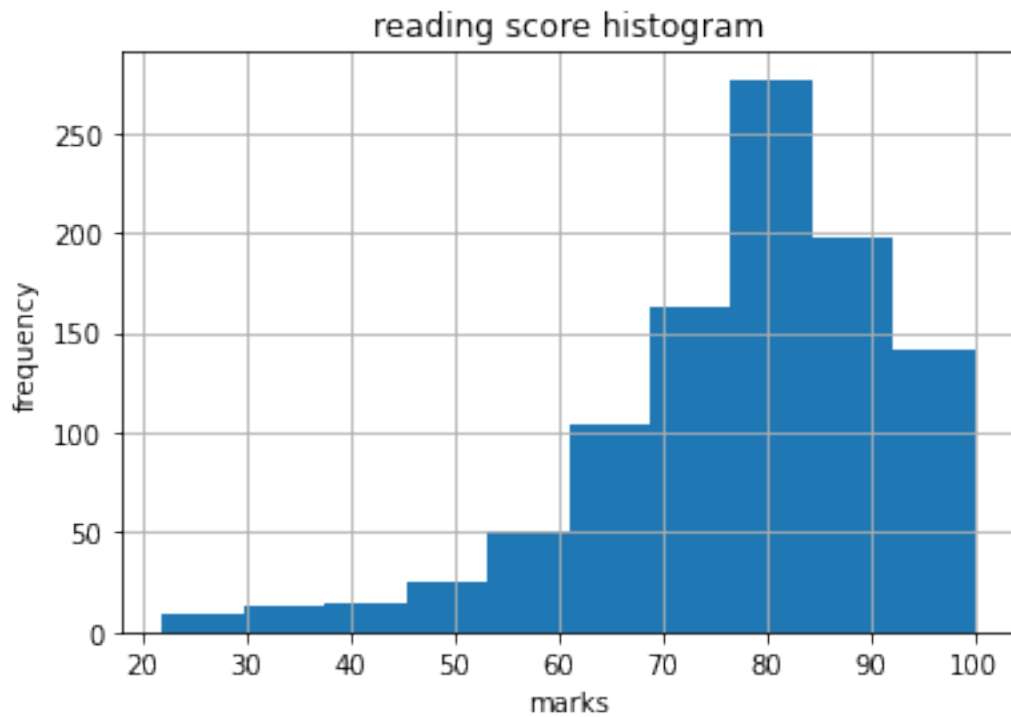
plot of the marks scored by 100 students

```
readingscores = df['reading score']
readingscores.plot()
```

<AxesSubplot:>



```
readingscores.hist()  
plt.ylabel("frequency")  
plt.xlabel("marks")  
plt.title('reading score histogram')  
Text(0.5, 1.0, 'reading score histogram')
```



```
scipy.stats.probplot(readingscores.values, dist="norm", plot=plt)

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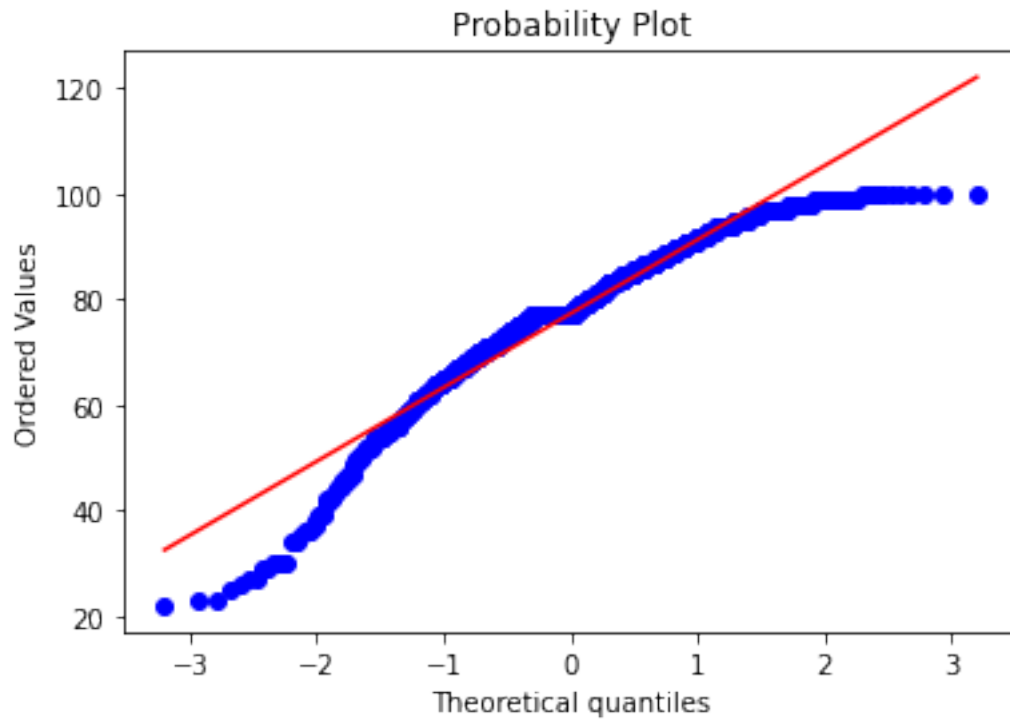
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Excluding Outliers the distribution is normal  $f(x) = 1/(\text{sd}\sqrt{2\pi})e^{(-1/2(x-\text{mean})/\text{sd})^2}$   
 mean median mode are equal distribution can be described by the mean and median

generate grades of students

```
def grade(x):
    if(x>90 and x<100):
        return 'S'
    elif(x>80):
        return 'A'
    elif(x>70):
        return 'B'
    elif(x>60):
        return 'C'
    elif(x>40):
        return 'D'
    else:
        return 'F'

df['grade']=df['percentage'].apply(lambda x:grade(x))
df['grade'].head()

0    D
1    D
2    F
3    D
```

4 C

Name: grade, dtype: object

```
#replacing some high school with high school and some college with college
df['parental level of education'].replace('some high school','high school',inplace=True)
df['parental level of education'].replace('some college', 'college', inplace=True)
parent_percentage_relation = df[['parental level of education', 'percentage', 'gender']]
parent_percentage_relation.dropna(subset=['gender'],inplace=True)
parent_percentage_relation['gender'] = parent_percentage_relation['gender'].str.lower()
parent_percentage_relation_male = parent_percentage_relation[parent_percentage_relation['gender']=='male']
parent_percentage_relation_female = parent_percentage_relation[parent_percentage_relation['gender'] == 'female']
parent_percentage_relation_male = parent_percentage_relation_male.groupby(['parental level of education']).mean()
parent_percentage_relation_female = parent_percentage_relation_female.groupby(['parental level of education']).mean()

ppgr = pd.DataFrame()
# ppgr['parental level of education'] =parent_percentage_relation_female['parental level of education'].values
ppgr['male'] = parent_percentage_relation_male['percentage'].values
ppgr['female']=parent_percentage_relation_female['percentage'].values

ppgr['parental level of education']=(parent_percentage_relation_male.index)
ppgr.plot(x='parental level of education',y=["male", "female"], kind="bar")
```

C:\Users\prach\AppData\Local\Temp\ipykernel\_10724\1648020781.py:7:

SettingWithCopyWarning:

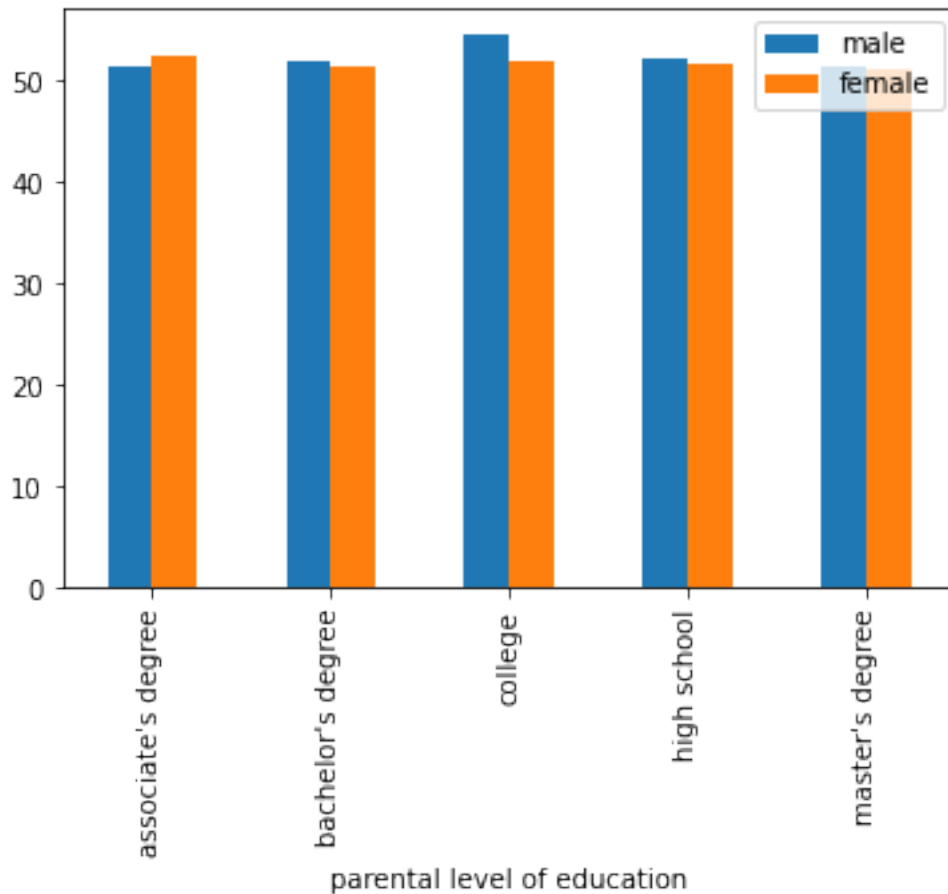
A value is trying to be set on a copy of a slice from a DataFrame.  
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation:

[https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
parent_percentage_relation['gender'] =
parent_percentage_relation['gender'].str.lower()
```

<AxesSubplot:xlabel='parental level of education'>



Task questions

```
set(df['race'].values) # checking if there are impossible values
df['race'].isnull().sum() # no nan values

simrstudents = df.sample(n=100)
stratstudents = df.groupby('race', group_keys=False).apply(lambda x:
x.sample(frac = 0.0997))
simmathmean = simrstudents['math score'].mean()
stratmathmean = stratstudents['math score'].mean()

simerror = mathscoremean - simmathmean
straterror = stratmathmean - mathscoremean
```

Sampling Error is lower in case of stratified random sampling

df

	gender	race	parental level of education	lunch
0	female	group B	bachelor's degree	standard

1	female	group C	high school	free/reduced
2	male	group C	master's degree	free/reduced
3	female	group C	high school	free/reduced
4	female	group C	associate's degree	free/reduced
..	...	...	...	...
995	female	group A	high school	standard
996	male	group B	bachelor's degree	standard
997	female	group C	associate's degree	standard
998	female	group D	bachelor's degree	free/reduced
999	female	group C	master's degree	free/reduced

	test preparation course	math score	reading score	writing score
\				
0	none	99.000000	85.00000	88.000000
1	none	96.000000	77.35078	77.903955
2	none	83.587464	29.00000	77.903955
3	completed	74.000000	70.00000	58.000000
4	none	83.587464	98.00000	89.000000
..	...	...	...	...
995	none	83.587464	77.35078	77.903955
996	none	89.000000	68.00000	69.000000
997	none	86.000000	84.00000	79.000000
998	none	95.000000	91.00000	91.000000
999	completed	83.587464	99.00000	100.000000

	percentage	grades	grade
0	57.666667	D	D
1	51.751578	D	D
2	35.634652	F	F
3	42.666667	D	D
4	62.333333	C	C
..	...	...	...
995	51.751578	D	D
996	45.666667	D	D
997	54.333333	D	D
998	60.666667	C	C
999	66.333333	C	C

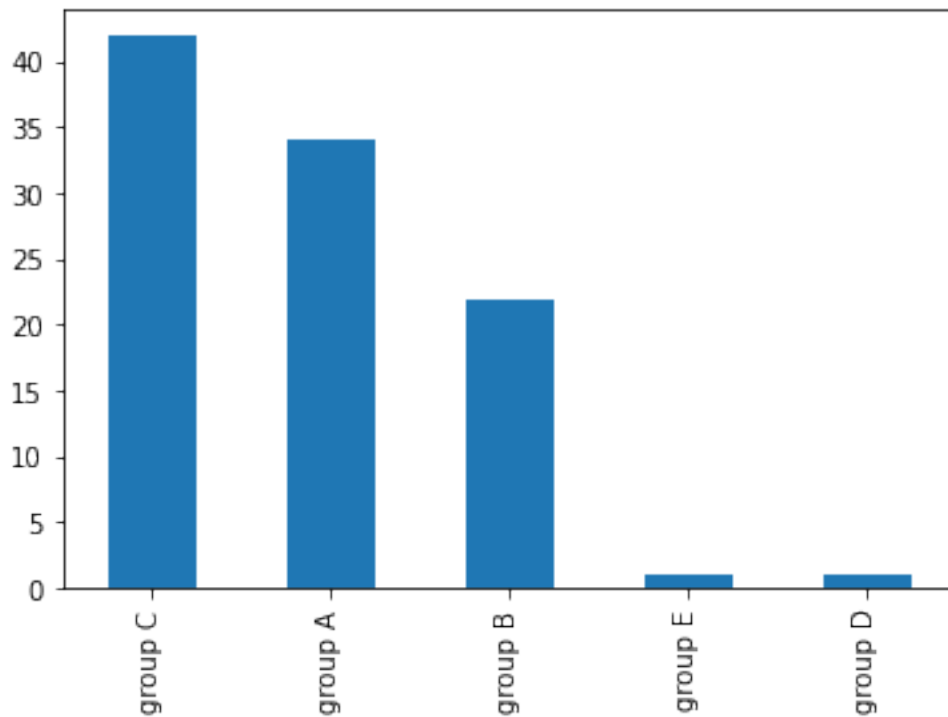
[995 rows x 11 columns]

Plotting bar graph of race distribution in samples

Simple random sampling (Bar graph with respect to race)

```
simrstudents['race'].value_counts().plot(kind='bar')
```

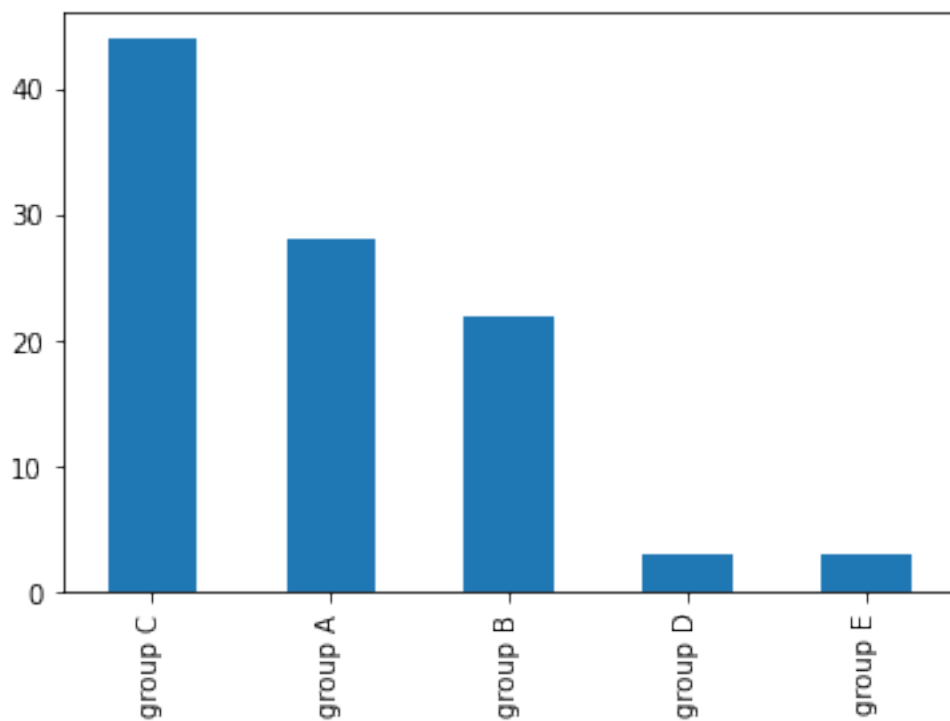
<AxesSubplot:>



Stratified random sampling (Bar graph with respect to race)

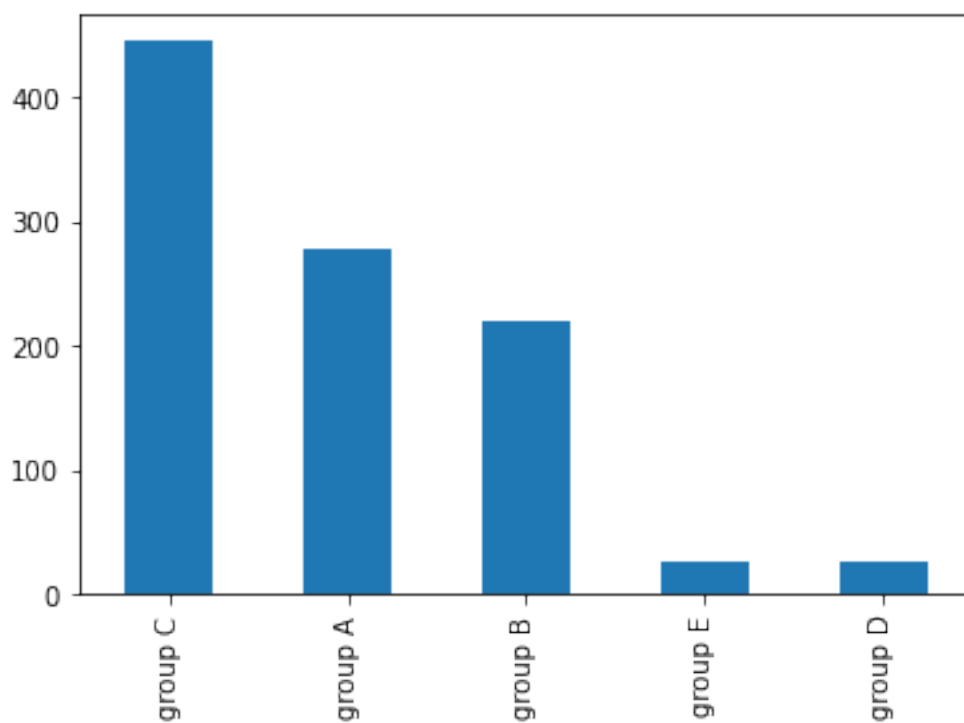
```
stratstudents['race'].value_counts().plot(kind='bar')
```

<AxesSubplot:>



```
df['race'].value_counts().plot(kind='bar')
```

<AxesSubplot:>



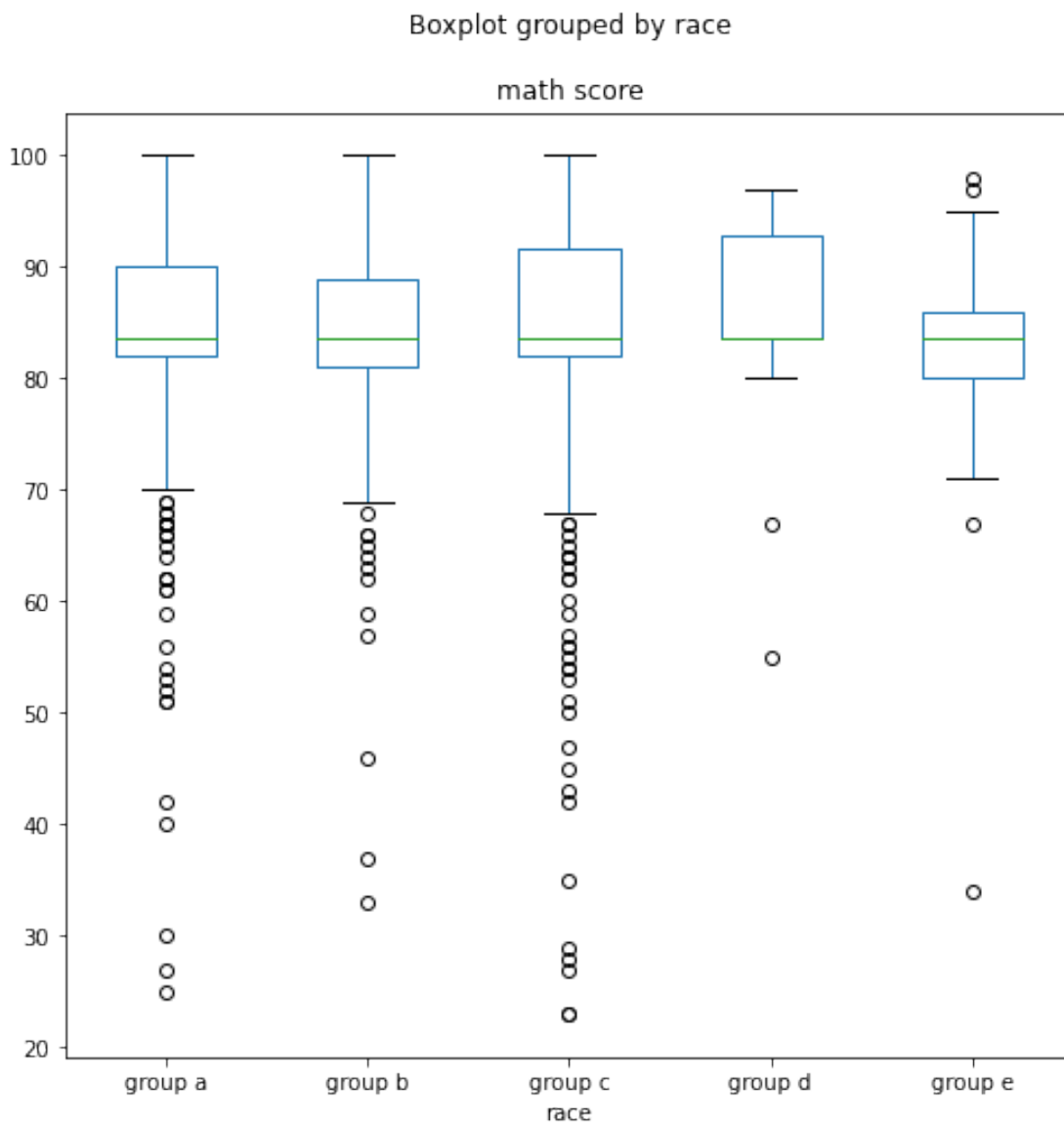


The graphs for simple and stratified random sampling are very similar to that of the Population However the stratified random sample has better representation of the distribution of race in the actual population

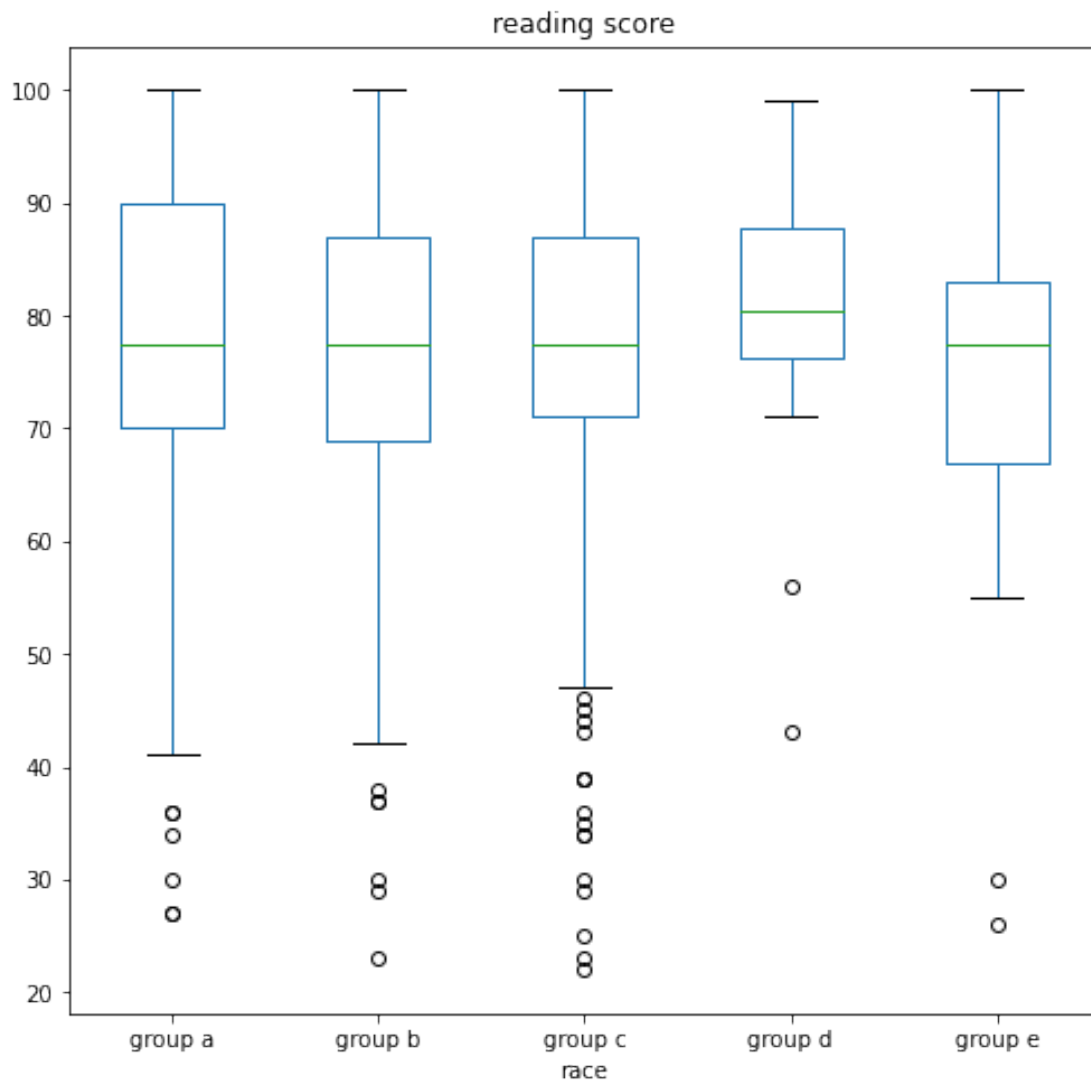
```
df['race'] = df['race'].str.lower()
df['race'].dropna()
```

```
df.boxplot(by='race',column=['math score'],grid=False,figsize=(8,8))
df.boxplot(by='race', column=['reading score'], grid=False,
figsize=(8, 8))
df.boxplot(by='race',column=['writing
score'],grid=False,figsize=(8,8))
```

```
<AxesSubplot:title={'center':'writing score'}, xlabel='race'>
```



Boxplot grouped by race



Boxplot grouped by race

