

To import necessary modules

```
In [1]: 1 import pandas as pd
2 import numpy as np
3 import warnings
4 warnings.filterwarnings("ignore")

In [2]: 1 test=pd.read_csv(r"K:\Desktop\NIIT\tables\DS1_C4_S4_Test_Scores_Data_Practice.csv")
2 non_unique=test.columns[test.columns!="student_id"]

In [3]: 1 school=test[non_unique]
```

Task1: To help find all the duplicate records after thier first occurence

```
In [4]: 1 school[school.duplicated(keep="first")]
```

Out[4]:

	school	school_setting	school_type	classroom	teaching_method	gender		lunch	pretest	posttest
	4	ANKYI	Urban	Non-public	6OL	Standard	Male	Does not qualify	64.0	76.0
	19	ANKYI	Urban	Non-public	6OL	Standard	Male	Does not qualify	64.0	73.0
	38	ANKYI	Urban	Non-public	ZNS	Standard	Female	Does not qualify	59.0	69.0
	50	CCAAW	Suburban	Non-public	2B1	Experimental	Male	Qualifies for reduced/free lunch	59.0	74.0
	106	CCAAW	Suburban	Non-public	PGK	Standard	Female	Does not qualify	73.0	80.0

	2058	ZOWMK	Urban	Public	Q0E	Experimental	Female	Qualifies for reduced/free lunch	38.0	53.0
	2067	ZOWMK	Urban	Public	Q0E	Experimental	Male	Qualifies for reduced/free lunch	37.0	53.0
	2070	ZOWMK	Urban	Public	Q0E	Experimental	Female	Qualifies for reduced/free lunch	38.0	53.0
	2080	ZOWMK	Urban	Public	QA2	Standard	Female	Qualifies for reduced/free lunch	45.0	51.0
	2128	ZOWMK	Urban	Public	ZBH	Standard	Female	Does not qualify	39.0	55.0

79 rows × 9 columns

Task2: To find all duplicates of school type teaching method pre test and post test parameters

```
In [5]: 1 school[school.duplicated(["school_type", "teaching_method", "pretest", "posttest"],keep="first")]
```

Out[5]:

	school	school_setting	school_type	classroom	teaching_method	gender		lunch	pretest	posttest
	4	ANKYI	Urban	Non-public	6OL	Standard	Male	Does not qualify	64.0	76.0
	14	ANKYI	Urban	Non-public	6OL	Standard	Female	Does not qualify	64.0	77.0
	19	ANKYI	Urban	Non-public	6OL	Standard	Male	Does not qualify	64.0	73.0
	26	ANKYI	Urban	Non-public	ZNS	Standard	Female	Does not qualify	60.0	70.0
	35	ANKYI	Urban	Non-public	ZNS	Standard	Female	Does not qualify	60.0	68.0

	2128	ZOWMK	Urban	Public	ZBH	Standard	Female	Does not qualify	39.0	55.0
	2129	ZOWMK	Urban	Public	ZBH	Standard	Female	Qualifies for reduced/free lunch	38.0	46.0
	2130	ZOWMK	Urban	Public	ZBH	Standard	Female	Qualifies for reduced/free lunch	45.0	51.0
	2131	ZOWMK	Urban	Public	ZBH	Standard	Male	Qualifies for reduced/free lunch	46.0	53.0
	2132	ZOWMK	Urban	Public	ZBH	Standard	Male	Qualifies for reduced/free lunch	41.0	48.0

903 rows × 9 columns

Task3: To find average pretest score for all schools and highest pretest score

```
In [6]: 1 avg=school.groupby(["school"]).mean()["pretest"]
2 print(avg)
3 print()
4 print("The school with the highest average pretest is = ",avg[avg==avg.max()])

school
ANKYI    61.341463
CCAAW    64.623853
CIMBB    65.067568
CUQAM    53.925234
DNQDD    54.327869
FBUMG    62.891304
GJJHK    53.194915
GOKXL    50.796875
GOOBU    38.248408
IDGFP    75.202128
KFZMY    41.865385
KZKKE    37.261261
LAYPA    62.035088
OJOB    56.197531
QOQTS    52.527027
UAGPU    62.390805
UKPGS    78.453125
UUUQX    67.253012
VHDHF    52.666667
VKWQH    52.060000
VVTVA    35.955752
ZMNYA    68.130435
ZOWMK    41.572650
Name: pretest, dtype: float64

The school with the highest average pretest is = school
UKPGS    78.453125
Name: pretest, dtype: float64
```

Task4: To display the location wise and gender wise maximum

```
In [7]: 1 school.loc[:,["school_setting", "gender"]].value_counts()[::-1]

Out[7]: school_setting  gender
Urban                Female    456
                   Male      444
dtype: int64
```

Task5: To display the minimum and maximum post test scores for all the schools and also show the top five post test scores Top 5

```
In [8]: posttest_sc=pd.pivot_table(school,index="school",values=["posttest"],aggfunc=["min","max"])
print("The minimum and maximum posttest scores by all schools")
posttest_sc
```

The minimum and maximum posttest scores by all schools

Out[8]:

	min	max
	posttest	posttest
school		
ANKYI	63.0	79.0
CCAAW	67.0	91.0
CIMBB	64.0	88.0
CUQAM	56.0	76.0
DNQDD	49.0	79.0
FBUMG	68.0	88.0
GJJHK	49.0	85.0
GOKXL	48.0	77.0
GOOBU	32.0	64.0
IDGFP	74.0	100.0
KFZMY	44.0	67.0
KZKKE	36.0	62.0
LAYPA	63.0	84.0
OJOBU	50.0	84.0
QOQTS	51.0	85.0
UAGPU	62.0	83.0
UKPGS	82.0	99.0
UUUQX	62.0	91.0
VHDHF	52.0	77.0
VKWQH	48.0	82.0
VVTVA	39.0	62.0
ZMNYA	66.0	95.0
ZOWMK	43.0	63.0

```
In [9]: print("The top 5 posttest scores ")
m=pd.pivot_table(school,index="school",values=["posttest"],aggfunc=["max"]).sort_values(by=[('max', 'posttest')],ascending=False).iloc[0:5,:]
```

The top 5 posttest scores

Out[9]:

	max
	posttest
school	
IDGFP	100.0
UKPGS	99.0
ZMNYA	95.0
UUUQX	91.0
CCAAW	91.0

Task6: To display school wise averages and sum of pre test and top 5 pre test scores

```
In [10]: print("The school wise average of pretest and posttest")
print(school.groupby("school").mean())
```

The school wise average of pretest and posttest

	pretest	posttest
school		
ANKYI	61.341463	71.390244
CCAAW	64.623853	78.110092
CIMBB	65.067568	76.945946
CUQAM	53.925234	65.560748
DNQDD	54.327869	66.565574
FBUMG	62.891304	78.608696
GJJHK	53.194915	65.025424
GOKXL	50.796875	64.953125
GOOBU	38.248408	49.643312
IDGFP	75.202128	87.223404
KFZMY	41.865385	54.576923
KZKKE	37.261261	47.918919
LAYPA	62.035088	73.571429
OJOBU	56.197531	67.814815
QOQTS	52.527027	64.671141
UAGPU	62.390805	71.873563
UKPGS	78.453125	91.188976
UUUQX	67.253012	79.261905
VHDHF	52.666667	66.843137
VKWQH	52.060000	64.820000
VVTVA	35.955752	49.203540
ZMNYA	68.130435	81.608696
ZOWMK	41.572650	52.905983

In [11]:

```
print("The sum of pretest scores : ")
print(school.groupby("school").sum()["pretest"])
print()
print("The top 5 average pretest scores = ")
print(school.groupby("school")["pretest"].mean().sort_values(ascending=False).head())
```

The sum of pretest scores :

school

ANKYI	2515.0
CCAAW	7044.0
CIMBB	4815.0
CUQAM	5770.0
DNQDD	6628.0
FBUMG	2893.0
GJJHK	6277.0
GOKXL	3251.0
GOOBU	6005.0
IDGFP	7069.0
KFZMY	2177.0
KZKKE	4136.0
LAYPA	3536.0
OJOBV	4552.0
QOQTS	7774.0
UAGPU	5428.0
UKPGS	10042.0
UUUQX	5582.0
VHDHF	2686.0
VKWQH	5206.0
VVTVA	4063.0
ZMNYA	4701.0
ZOWMK	4864.0

Name: pretest, dtype: float64

The top 5 average pretest scores =

school

UKPGS	78.453125
IDGFP	75.202128
ZMNYA	68.130435
UUUQX	67.253012
CIMBB	65.067568

Name: pretest, dtype: float64

Task7: The school FBUMG with type of 5LQ classroom wishes to offer scholarships

In [12]:

```
FBU=test[(test.school=="FBUMG") & (test.classroom=="5LQ")].loc[:,["student_id","posttest"]]
FBU["Scholarship"]=(FBU.posttest/100)*10000
FBU
```

Out[12]:

	student_id	posttest	Scholarship
467	04DG5	84.0	8400.0
468	20M2D	85.0	8500.0
469	39KCV	81.0	8100.0
470	5Z1B6	88.0	8800.0
471	6TLU8	81.0	8100.0
472	ATQQJ	82.0	8200.0
473	AYEU1	79.0	7900.0
474	B9FSU	83.0	8300.0
475	EV13K	83.0	8300.0
476	I5H37	83.0	8300.0
477	JC5I9	80.0	8000.0
478	JPE2J	81.0	8100.0
479	MFBYU	82.0	8200.0
480	O144X	86.0	8600.0
481	OGKP3	83.0	8300.0
482	UMFI7	73.0	7300.0
483	V1DNJ	76.0	7600.0
484	YRN9S	87.0	8700.0

Task8: To replace does not qualify and qualifies with reduced free lunch with NO and YES

In [13]:

```
school.lunch.replace(to_replace=["Does not qualify","Qualifies for reduced/free lunch"],value=["No","Yes"],inplace=True)
school.lunch
```

Out[13]:

0	No
1	No
2	No
3	No
4	No
...	
2128	No
2129	Yes
2130	Yes
2131	Yes
2132	Yes

Name: lunch, Length: 2133, dtype: object

Task9: To identify count of teaching methods by locations

In [14]:

```
pd.crosstab(school.school_setting,school.teaching_method)
```

Out[14]:

teaching_method	Experimental	Standard
school_setting		
Rural	201	309
Suburban	284	433
Urban	275	631

In [15]:

```
school.value_counts().groupby(["school_setting","teaching_method"]).sum()
```

Out[15]:

school_setting	teaching_method	
Rural	Experimental	199
	Standard	307
Suburban	Experimental	283
	Standard	431
Urban	Experimental	269
	Standard	626

dtype: int64

Task10: To replace the pre test scores of male students with new value of 135 for male students who have post test less than 39

```
In [16]: index=school[(school.gender=="Male")&(school.posttest<39)][ "pretest"].index
school.pretest[index.values]=135
school.pretest[index.values]
school[(school.gender=="Male")&(school.posttest<39)]
```

Out[16]:

	school	school_setting	school_type	classroom	teaching_method	gender	lunch	pretest	posttest
742	GOOBU	Urban	Public	HKF	Standard	Male	Yes	135.0	38.0
745	GOOBU	Urban	Public	HKF	Standard	Male	Yes	135.0	38.0
751	GOOBU	Urban	Public	HKF	Standard	Male	Yes	135.0	34.0
1023	KZKKE	Rural	Public	3D0	Standard	Male	Yes	135.0	37.0
1081	KZKKE	Rural	Public	QTU	Standard	Male	Yes	135.0	38.0
1089	KZKKE	Rural	Public	QTU	Standard	Male	Yes	135.0	37.0

Task11: To compare average posttest scores with sub urban and urban male students

```
In [17]: comp=pd.pivot_table(school,index="school_setting",values=["posttest"],aggfunc=["mean"]).iloc[0:2,:]
print(comp)
print()
print("relative percentage = ",round((((comp.values[1][0]-comp.values[0][0])/comp.values[0][0])*100,2), "%")
```

	mean
posttest	
school_setting	
Rural	64.039293
Suburban	76.020950

relative percentage = 18.71 %

Task12: To help replace missing values with most recurring values

```
In [18]: print(school.isnull().sum())

school.gender=school.gender.fillna("Male")
school.pretest=school.pretest.fillna(school.pretest.mean())
school.posttest=school.posttest.fillna(school.posttest.mean())
print("After replacing ")
print()
print(school.isnull().sum())
```

school	0
school_setting	0
school_type	0
classroom	0
teaching_method	0
gender	10
lunch	0
pretest	4
posttest	4
dtype: int64	
After replacing	
school	0
school_setting	0
school_type	0
classroom	0
teaching_method	0
gender	0
lunch	0
pretest	0
posttest	0
dtype: int64	