# REGRESSION WITH DUMMY VARIABLES

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### Introduction

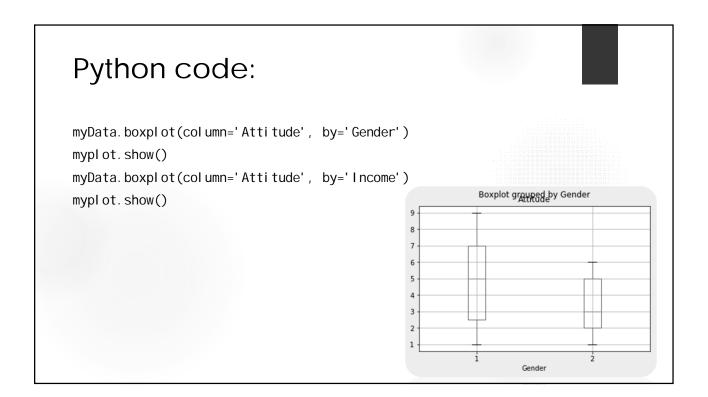
- ▶ When Xs are not numeric but nominal
- ► Each nominal or categorical variable is converted into dummy variables
- ▶ Dummy Variables will take values 0 or 1
- ▶ Number of dummy variables for one X variable is equal to number of distinct values of that variable 1

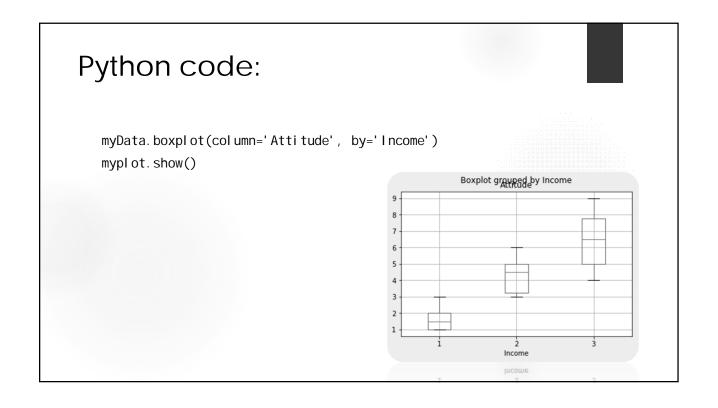
### **Exercise:**

- ► A study was conducted to measure the effect of gender and income on attitude towards vocation. Data was collected from 30 respondents and is given in vocation\_dummy\_reg.csv file.
- ► Attitude towards vocation is measured on a 9 point scale. Gender is coded as male =1 and female =2
- ▶ Income is coded as low=1, medium=2 and high =3
- Develop a model for attitude towards vocation in terms of gender and income

## Python code:

import pandas as mypanda
from scipy import stats
import matplotlib.pyplot as myplot
from statsmodels.formula.api import ols
myData=mypanda.read\_csv('vocation\_dummy\_Reg.csv')
myData
gender=myData.Gender
income=myData.Income
attitude=myData.Attitude

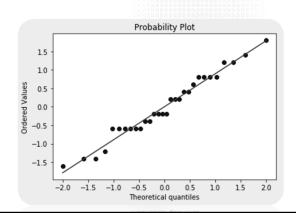




# Python code:

```
mymodel.summary()
pred=mymodel.predict()
pred
res=attitude-pred
stats.probplot(res, plot=myplot)
myplot.show()
```

mymodel = ol s('attitude ~ C(gender) + C(income)', myData). fit()



# Python code:

stats.normal test(res)
Out[1 Normal testResult(stat)

Out[] Normal testResul t(stati sti c=0. 52111989611555032, pval ue=0. 7706199578215539)

from statsmodels.stats.anova import anova\_I  $\mbox{\it m}$ 

anova\_table = anova\_lm(mymodel)
anova\_table

		df	sum_sq	mean_sq	F	PR(>F)
-	C(gender)	1.0	19.200000	19.200000	22.690909	6.274380e-05
(	C(income)	2.0	116.266667	58.133333	68.703030	4.189551e-11
	Residual	26.0	22.000000	0.846154	NaN	NaN

