

Old Dataset(Staircase)

Importing Jury data csv file to data and renaming the columns ¶

In [1]:

```
# coding: utf-8
import pandas as pd
import numpy as np

data = pd.read_csv('jury_data.csv', encoding= 'ISO-8859-1', skiprows=[0,2])
data.rename(columns={"Was defendant Mesa Management negligent?": "Mesa_Negligent",
                    "Was Mesa Management's negligence a substantial factor in causing harm to Mackenzie Dunn?": "Liability",
                    "What are the total damages that you find that MacKenzie Dunn sufferered?": "damages" ,
                    "What is your sex?": "gender",
                    "Please write your answer to the preceding damages question in words (quality check).": "damages_word",
                    "What percentage of responsibility for Mackenzie Dunn's injuries was each party responsible for? (Answers should add up to 100%) - Mesa Management Co": "Mesa_reponsible_percentage",
                    "Path": "Scenario",
                    "Was MacKenzie Dunn negligent?": "Dunn_negligent",
                    "Unnamed: 63": "perc_calc",
                    "Which of the following best describes your total household income?": "Income"
                    }, inplace=True)
#data['mm_perc'] = np.where(data['Mesa_reponsible_percentage']>=1, data['perc_calc'], data['Mesa_reponsible_percentage'])
data['mm_perc'] = data['Mesa_reponsible_percentage']
req_data = pd.DataFrame(data[["Mesa_Negligent", "damages", "Liability", "gender", "damages_word", "Scenario", "Dunn_negligent", "perc_calc", "Start Date", "End Date", "mm_perc", "Income"]])

req_data['Liability'] = req_data['Liability'].map({'Yes': 1, 'No': 0})

print(req_data.columns)

Index(['Mesa_Negligent', 'damages', 'Liability', 'gender', 'damages_word', 'Scenario', 'Dunn_negligent', 'perc_calc', 'Start Date', 'End Date', 'mm_perc', 'Income'],
      dtype='object')
```

In [2]:

```
data['mm_perc'].unique()
```

Out[2]:

```
array([ nan,  0.65,  65.,  0.6,  0.1,
        0.5,  0.7,  0.85,  0.45,  0.,
        0.75,  0.51,  0.9,  0.3,  0.2,
        0.4,  0.49526316])
```

Replacing Null values and changing the data types of required columns

In [3]:

```
print(pd.isnull(req_data).any())
print(pd.isnull(req_data['Scenario']).any())
req_data = req_data[np.isfinite(data['Scenario'])]
print(pd.isnull(req_data['Scenario']).any())
req_data['damages'].fillna(0,inplace=True)
req_data['damages_word'].fillna(0,inplace=True)
req_data['mm_perc'].fillna(1,inplace=True)
req_data['perc_calc'].fillna(0,inplace=True)
#Dropping the last two rows which has null values
#data[pd.isnull(data['Path'])]
#data['Path']=data.Path.dropna(inplace= True)
#data[pd.isnull(data['Path'])]
print(pd.isnull(req_data).any())

#Changing data types of columns
req_data['End Date'] = pd.to_datetime(data['End Date'])
req_data['Start Date'] = pd.to_datetime(data['Start Date'])
req_data['Scenario']= req_data.Scenario.astype(int)
req_data['Liability']= req_data.Liability.astype(int)
req_data.dtypes

# Getting the id of the column
data.columns.get_loc("Liability")
```

```
Mesa_Negligent      True
damages             True
Liability           True
gender              True
damages_word        True
Scenario            True
Dunn_negligent      True
perc_calc           True
Start Date          False
End Date            False
mm_perc             True
Income              True
```

```
dtype: bool
```

```
True
```

```
False
```

```
Mesa_Negligent      False
damages             False
Liability           False
gender              False
damages_word        False
Scenario            False
Dunn_negligent      True
perc_calc           False
Start Date          False
End Date            False
mm_perc             False
Income              False
```

```
dtype: bool
```

Out[3]:

55

Cleaning Damages and perc_calc column

In [4]:

```
print(req_data.isnull().any())
req_data['damages'] = req_data['damages'].str.replace(',', '')
req_data['perc_calc'] = req_data['perc_calc'].str.replace('$', '')
req_data['perc_calc'] = req_data['perc_calc'].str.replace('.', '', '')
req_data['perc_calc'] = req_data['perc_calc'].str.replace('-', '')
req_data['perc_calc'] = req_data['perc_calc'].str.replace(" ", '')

#req_data['mm_perc'] = req_data['mm_perc'].str.replace("$", '')
#req_data['mm_perc'] = req_data['mm_perc'].str.replace(",", '')
#req_data['mm_perc'] = req_data['mm_perc'].str.replace(" ", '')
#req_data.damages=pd.to_numeric(req_data['damages'].str.replace(',', ''))
#req_data.perc_calc=pd.to_numeric(req_data.perc_calc)

req_data.damages=pd.to_numeric(req_data['damages'])
req_data.perc_calc=pd.to_numeric(req_data.perc_calc)
req_data['damages'].fillna(0,inplace=True)
req_data['mm_perc'].fillna(1,inplace=True)
req_data['perc_calc'].fillna(0,inplace=True)
#print(req_data.damages)
print(req_data.isnull().any())

print(req_data[pd.isnull(req_data['Dunn_negligent'])])
```

```

Mesa_Negligent      False
damages              False
Liability            False
gender               False
damages_word         False
Scenario             False
Dunn_negligent       True
perc_calc            False
Start Date           False
End Date             False
mm_perc              False
Income               False
dtype: bool
Mesa_Negligent      False
damages              False
Liability            False
gender               False
damages_word         False
Scenario             False
Dunn_negligent       True
perc_calc            False
Start Date           False
End Date             False
mm_perc              False
Income               False
dtype: bool

```

	Mesa_Negligent	damages	Liability	gender	damages_word	Scenari
0	No	0.0	0	Female	0	
1						
1	No	0.0	0	Male	0	
1						
2	No	0.0	0	Male	0	
1						
3	No	0.0	0	Male	0	
1						
4	No	0.0	0	Male	0	
1						
5	No	0.0	0	Male	0	
1						
6	No	0.0	0	Male	0	
1						
7	No	0.0	0	Female	0	
1						
8	No	0.0	0	Male	0	
1						
9	No	0.0	0	Female	0	
1						
10	No	0.0	0	Male	0	
1						
11	No	0.0	0	Female	0	
1						
12	No	0.0	0	Male	0	
1						
13	No	0.0	0	Female	0	
1						
14	No	0.0	0	Male	0	
1						
15	No	0.0	0	Male	0	
1						
16	No	0.0	0	Male	0	

1					
17	No	0.0	0	Male	0
1					
18	No	0.0	0	Male	0
1					
19	No	0.0	0	Male	0
1					
20	No	0.0	0	Male	0
1					
21	No	0.0	0	Female	0
1					
22	No	0.0	0	Female	0
1					
23	No	0.0	0	Female	0
1					
24	No	0.0	0	Male	0
1					
25	No	0.0	0	Male	0
1					
26	No	0.0	0	Male	0
1					
27	No	0.0	0	Male	0
1					
28	No	0.0	0	Female	0
1					
29	No	0.0	0	Male	0
1					
..
...					
397	Yes	0.0	0	Female	0
2					
398	Yes	0.0	0	Male	0
2					
399	Yes	0.0	0	Female	0
2					
400	Yes	0.0	0	Female	0
2					
401	Yes	0.0	0	Female	0
2					
402	Yes	0.0	0	Female	0
2					
403	Yes	0.0	0	Female	0
2					
404	Yes	0.0	0	Female	0
2					
405	Yes	0.0	0	Female	0
2					
406	Yes	0.0	0	Female	0
2					
407	Yes	0.0	0	Male	0
3					
408	Yes	0.0	0	Female	0
3					
409	Yes	0.0	0	Male	0
3					
410	Yes	0.0	0	Female	0
4					
411	Yes	0.0	0	Male	0
4					
412	Yes	0.0	0	Male	0
4					

413	Yes	0.0	0	Male	0
4					
414	Yes	0.0	0	Male	0
4					
415	Yes	0.0	0	Male	0
4					
416	Yes	0.0	0	Male	0
4					
417	Yes	0.0	0	Female	0
5					
418	Yes	0.0	0	Male	0
5					
419	Yes	0.0	0	Female	0
5					
420	Yes	0.0	0	Female	0
5					
421	Yes	0.0	0	Male	0
5					
422	Yes	0.0	0	Female	0
5					
423	Yes	0.0	0	Female	0
5					
424	Yes	0.0	0	Female	0
5					
425	Yes	0.0	0	Female	0
5					
426	Yes	0.0	0	Male	0
5					

	Dunn_negligent	perc_calc	Start Date	End Date
e \				
0	NaN	0.0	2017-09-29 13:58:00	2017-09-29 14:16:00
0				
1	NaN	0.0	2017-09-29 14:00:00	2017-09-29 14:18:00
0				
2	NaN	0.0	2017-09-29 13:57:00	2017-09-29 14:19:00
0				
3	NaN	0.0	2017-09-29 14:01:00	2017-09-29 14:21:00
0				
4	NaN	0.0	2017-09-29 14:04:00	2017-09-29 14:23:00
0				
5	NaN	0.0	2017-09-29 14:06:00	2017-09-29 14:25:00
0				
6	NaN	0.0	2017-09-29 14:06:00	2017-09-29 14:26:00
0				
7	NaN	0.0	2017-09-29 14:16:00	2017-09-29 14:35:00
0				
8	NaN	0.0	2017-09-29 14:27:00	2017-09-29 14:49:00
0				
9	NaN	0.0	2017-09-29 14:44:00	2017-09-29 15:02:00
0				
10	NaN	0.0	2017-09-29 15:09:00	2017-09-29 15:27:00
0				
11	NaN	0.0	2017-09-29 15:55:00	2017-09-29 16:16:00
0				
12	NaN	0.0	2017-09-29 18:28:00	2017-09-29 18:51:00
0				
13	NaN	0.0	2017-09-29 19:54:00	2017-09-29 20:14:00
0				
14	NaN	0.0	2017-09-29 20:56:00	2017-09-29 21:16:00
0				

150	NaN	0.0	2017-09-30	11:47:00	2017-09-30	12:04:0
160	NaN	0.0	2017-09-30	12:56:00	2017-09-30	13:13:0
170	NaN	0.0	2017-09-30	14:15:00	2017-09-30	14:26:0
180	NaN	0.0	2017-09-30	16:19:00	2017-09-30	16:38:0
190	NaN	0.0	2017-10-01	21:22:00	2017-10-01	21:40:0
200	NaN	0.0	2017-10-01	21:47:00	2017-10-01	22:07:0
210	NaN	0.0	2017-10-02	02:43:00	2017-10-02	03:02:0
220	NaN	0.0	2017-10-02	06:47:00	2017-10-02	07:11:0
230	NaN	0.0	2017-10-02	07:12:00	2017-10-02	07:31:0
240	NaN	0.0	2017-10-02	07:21:00	2017-10-02	07:38:0
250	NaN	0.0	2017-10-02	07:57:00	2017-10-02	08:15:0
260	NaN	0.0	2017-10-02	08:31:00	2017-10-02	08:50:0
270	NaN	0.0	2017-10-02	09:01:00	2017-10-02	09:19:0
280	NaN	0.0	2017-10-02	09:20:00	2017-10-02	09:43:0
290	NaN	0.0	2017-10-02	10:41:00	2017-10-02	10:59:0
..		
...						
3970	NaN	0.0	2017-09-29	14:00:00	2017-09-29	14:19:0
3980	NaN	0.0	2017-09-29	18:28:00	2017-09-29	18:48:0
3990	NaN	0.0	2017-09-30	11:49:00	2017-09-30	12:10:0
4000	NaN	0.0	2017-09-30	14:52:00	2017-09-30	15:12:0
4010	NaN	0.0	2017-10-01	10:25:00	2017-10-01	10:42:0
4020	NaN	0.0	2017-10-03	13:07:00	2017-10-03	13:26:0
4030	NaN	0.0	2017-10-05	18:01:00	2017-10-05	18:20:0
4040	NaN	0.0	2017-10-05	18:16:00	2017-10-05	18:35:0
4050	NaN	0.0	2017-10-05	18:17:00	2017-10-05	18:40:0
4060	NaN	0.0	2017-10-05	18:20:00	2017-10-05	18:49:0
4070	NaN	0.0	2017-09-30	20:10:00	2017-09-30	20:30:0
4080	NaN	0.0	2017-10-05	17:45:00	2017-10-05	18:12:0
4090	NaN	0.0	2017-10-05	18:03:00	2017-10-05	18:25:0
4100	NaN	0.0	2017-09-29	14:14:00	2017-09-29	14:44:0
411	NaN	0.0	2017-09-29	18:35:00	2017-09-29	18:58:0

0						
412	NaN	0.0	2017-10-01	04:32:00	2017-10-01	04:54:0
0						
413	NaN	0.0	2017-10-02	09:26:00	2017-10-02	09:48:0
0						
414	NaN	0.0	2017-10-05	17:32:00	2017-10-05	17:52:0
0						
415	NaN	0.0	2017-10-05	17:33:00	2017-10-05	17:54:0
0						
416	NaN	0.0	2017-10-05	17:39:00	2017-10-05	18:02:0
0						
417	NaN	0.0	2017-09-30	17:52:00	2017-09-30	18:11:0
0						
418	NaN	0.0	2017-10-03	07:48:00	2017-10-03	08:08:0
0						
419	NaN	0.0	2017-10-04	10:56:00	2017-10-04	11:17:0
0						
420	NaN	0.0	2017-10-05	17:25:00	2017-10-05	17:47:0
0						
421	NaN	0.0	2017-10-05	17:29:00	2017-10-05	17:50:0
0						
422	NaN	0.0	2017-10-05	17:25:00	2017-10-05	17:52:0
0						
423	NaN	0.0	2017-10-05	17:32:00	2017-10-05	17:55:0
0						
424	NaN	0.0	2017-10-05	17:32:00	2017-10-05	17:57:0
0						
425	NaN	0.0	2017-10-05	17:58:00	2017-10-05	18:19:0
0						
426	NaN	0.0	2017-10-05	18:10:00	2017-10-05	18:30:0
0						

	mm_perc	Income
0	1.0	\$10,000 to \$29,999
1	1.0	\$50,000 to \$99,999
2	1.0	\$30,000 to \$49,999
3	1.0	\$10,000 to \$29,999
4	1.0	\$30,000 to \$49,999
5	1.0	\$100,000 to \$199,999
6	1.0	\$10,000 to \$29,999
7	1.0	\$50,000 to \$99,999
8	1.0	\$100,000 to \$199,999
9	1.0	\$100,000 to \$199,999
10	1.0	Less than \$10,000
11	1.0	\$50,000 to \$99,999
12	1.0	\$50,000 to \$99,999
13	1.0	\$50,000 to \$99,999
14	1.0	\$10,000 to \$29,999
15	1.0	\$100,000 to \$199,999
16	1.0	\$50,000 to \$99,999
17	1.0	\$50,000 to \$99,999
18	1.0	\$10,000 to \$29,999
19	1.0	\$50,000 to \$99,999
20	1.0	\$10,000 to \$29,999
21	1.0	\$50,000 to \$99,999
22	1.0	\$30,000 to \$49,999
23	1.0	\$50,000 to \$99,999
24	1.0	\$50,000 to \$99,999
25	1.0	\$30,000 to \$49,999
26	1.0	\$30,000 to \$49,999
27	1.0	Less than \$10,000

28	1.0	\$100,000 to \$199,999
29	1.0	\$50,000 to \$99,999
..
397	1.0	Less than \$10,000
398	1.0	\$10,000 to \$29,999
399	1.0	\$30,000 to \$49,999
400	1.0	\$30,000 to \$49,999
401	1.0	\$10,000 to \$29,999
402	1.0	\$100,000 to \$199,999
403	1.0	\$50,000 to \$99,999
404	1.0	\$50,000 to \$99,999
405	1.0	\$100,000 to \$199,999
406	1.0	\$50,000 to \$99,999
407	1.0	\$100,000 to \$199,999
408	1.0	\$10,000 to \$29,999
409	1.0	\$30,000 to \$49,999
410	1.0	\$100,000 to \$199,999
411	1.0	\$10,000 to \$29,999
412	1.0	\$30,000 to \$49,999
413	1.0	\$30,000 to \$49,999
414	1.0	\$200,000 or more
415	1.0	\$50,000 to \$99,999
416	1.0	\$10,000 to \$29,999
417	1.0	\$10,000 to \$29,999
418	1.0	\$100,000 to \$199,999
419	1.0	\$10,000 to \$29,999
420	1.0	\$50,000 to \$99,999
421	1.0	\$10,000 to \$29,999
422	1.0	\$10,000 to \$29,999
423	1.0	\$100,000 to \$199,999
424	1.0	\$50,000 to \$99,999
425	1.0	\$10,000 to \$29,999
426	1.0	\$30,000 to \$49,999

[427 rows x 12 columns]

In [5]:

```
req_data['Income'] = req_data['Income'].str.replace(" ", '')
req_data['Income'] = req_data['Income'].str.replace('$', '')
req_data['Income'] = req_data['Income'].str.replace(',', '')
```

In [6]:

```
req_data['Income'].unique()
```

Out[6]:

```
array(['10000 to 29999', '50000 to 99999', '30000 to 49999',
      '100000 to 199999', 'Less than 10000', '200000 or more'],
      dtype=object)
```

In [7]:

```
req_data['Income']=req_data['Income'].replace(['Less than 10000','10000 to 29999',
'50000 to 99999','30000 to 49999','100000 to 199999',
'200000 or more','nan'],['1','2','3','4','5','6','0'])
```

In [8]:

```
req_data['Income'].unique()
```

Out[8]:

```
array(['2', '3', '4', '5', '1', '6'], dtype=object)
```

Calculating Discounted Damages

In [9]:

```
req_data['Discounted_damages']=req_data.damages*pd.to_numeric(req_data.mm_perc)
```

Replacing above 500k to 500k

In [10]:

```
req_data.loc[req_data.Discounted_damages>500000.0].shape  
req_data.Discounted_damages=np.where(((req_data.Discounted_damages>500000.0)),500000,req_data.Discounted_damages)
```

In [11]:

```
req_data.loc[req_data.Discounted_damages>500000.0].shape
```

Out[11]:

```
(0, 13)
```

EDA

Calculating Discounted Damages stats for case expected value

In [12]:

```

req_data['winrate_percentage']=req_data.Liability
req_data['Discounted_damages_mean']=req_data['Discounted_damages']
req_data['Discounted_damages_median']=req_data['Discounted_damages']
req_data['Discounted_damages_sd']=req_data['Discounted_damages']

winrate_damages_expected=req_data.groupby('Scenario').aggregate(
    {'winrate_percentage': np.mean, 'Discounted_damages_mean': np.mean, 'Discounted_damages_median': np.median, 'Discounted_damages_sd': np.std})

winrate_damages_expected['winrate_percentage']=winrate_damages_expected['winrate_percentage']*100.0
winrate_damages_expected

```

Out[12]:

	winrate_percentage	Discounted_damages_mean	Discounted_damages_me
Scenario			
1	47.263682	97124.378109	0.0
2	48.730964	98857.868020	0.0
3	68.817204	130739.247312	110000.0
4	60.294118	108101.715686	100000.0
5	57.591623	92761.780105	80000.0

Calculating Discounted Damages stats when plaintiff wins

In [21]:

```
#req_data['winrate_percentage']=np.mean(req_data.Juror_Response)
#print(req_data)
req_data['mm_perc'].fillna(1,inplace=True)

req_data['Discounted_damages_mean1']=req_data['Discounted_damages']
req_data['Discounted_damages_median1']=req_data['Discounted_damages']
req_data['Discounted_damages_sd1']=req_data['Discounted_damages']
#print(req_data.mm_perc)

winrate_damages_plaintiffwin=req_data.loc[(req_data['Mesa_Negligent']=='Yes') &
(req_data['Liability']==1)].groupby('Scenario').aggregate({'Discounted_damages_m
ean1': np.mean,'Discounted_damages_median1':np.median,'Discounted_damages_sd1':n
p.std})

#Here Liability is "Is substantial factor for causing plaintiff injuries"
winrate_damages_plaintiffwin
```

Out[21]:

	Discounted_damages_mean1	Discounted_damages_median1	Discounted_d
Scenario			
1	205494.736842	180000.0	111371.80147
2	202864.583333	180000.0	108895.76085
3	189980.468750	180000.0	102566.32964
4	179290.650407	180000.0	77900.192120
5	161068.181818	150000.0	81512.690442

Plotting graph for Liability vs Path

In [14]:

```
import matplotlib as mpl
import matplotlib.pyplot as plt
import seaborn as sns

sns.factorplot(x='Scenario', y='damages', kind='box',data=req_data)
```

Out[14]:

```
<seaborn.axisgrid.FacetGrid at 0x109903b70>
```

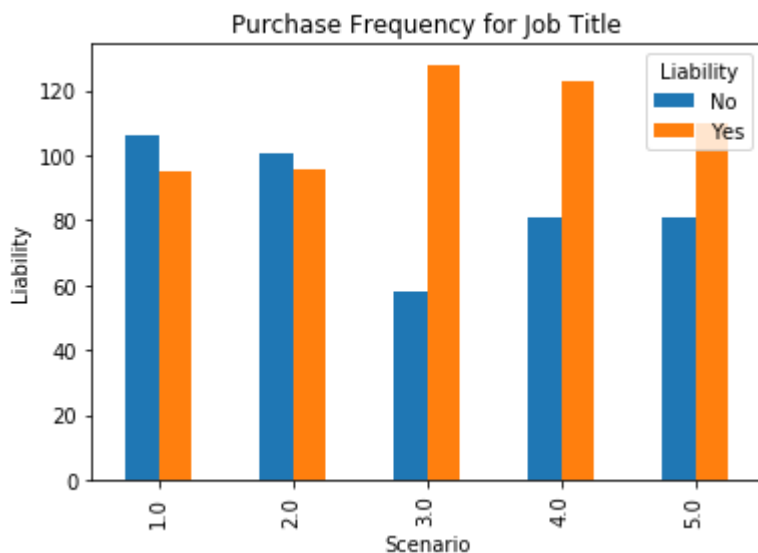
In [15]:

```
pd.crosstab(data.Scenario,data.Liability).plot(kind='bar')
plt.title('Purchase Frequency for Job Title')
plt.xlabel('Scenario')
plt.ylabel('Liability')
plt.savefig('Juror Response per each Scenario')

a = req_data['Scenario']
b = req_data['Liability']
pd.crosstab(a,b)
```

Out[15]:

Liability	0	1
Scenario		
1	106	95
2	101	96
3	58	128
4	81	123
5	81	110



From the graph we can see that

- Winrate increases when subsequent remedial measures introduced at scenario3.
- Winrate decreases slightly when limited jury instruction presented at Scenario4.
- Similarly winrate and damages decreases more when more explanation on limiting jury instruction provided at Scenario5. But not decreased as in Scenario2 (before limiting jury instructions)

Importing data to "old_data.csv" file

In [16]:

```
req_data.to_csv("old_data.csv",sep=',')
```