

# Earnings\_Visuals\_Based\_On\_College\_Major

February 28, 2020

```
[1]: import pandas as pd
import matplotlib as plt
%matplotlib inline

recent_grads = pd.read_csv('recent-grads.csv')
recent_grads.iloc[0]
recent_grads.head()
```

```
[1]:
```

	Rank	Major_code	Major	Total	\
0	1	2419	PETROLEUM ENGINEERING	2339.0	
1	2	2416	MINING AND MINERAL ENGINEERING	756.0	
2	3	2415	METALLURGICAL ENGINEERING	856.0	
3	4	2417	NAVAL ARCHITECTURE AND MARINE ENGINEERING	1258.0	
4	5	2405	CHEMICAL ENGINEERING	32260.0	

	Men	Women	Major_category	ShareWomen	Sample_size	Employed	...	\
0	2057.0	282.0	Engineering	0.120564	36	1976	...	
1	679.0	77.0	Engineering	0.101852	7	640	...	
2	725.0	131.0	Engineering	0.153037	3	648	...	
3	1123.0	135.0	Engineering	0.107313	16	758	...	
4	21239.0	11021.0	Engineering	0.341631	289	25694	...	

	Part_time	Full_time_year_round	Unemployed	Unemployment_rate	Median	\
0	270	1207	37	0.018381	110000	
1	170	388	85	0.117241	75000	
2	133	340	16	0.024096	73000	
3	150	692	40	0.050125	70000	
4	5180	16697	1672	0.061098	65000	

	P25th	P75th	College_jobs	Non_college_jobs	Low_wage_jobs
0	95000	125000	1534	364	193
1	55000	90000	350	257	50
2	50000	105000	456	176	0
3	43000	80000	529	102	0
4	50000	75000	18314	4440	972

[5 rows x 21 columns]

```
[2]: recent_grads.tail()
```

```
[2]:
```

	Rank	Major_code	Major	Total	Men	Women	\
168	169	3609	ZOOLOGY	8409.0	3050.0	5359.0	
169	170	5201	EDUCATIONAL PSYCHOLOGY	2854.0	522.0	2332.0	
170	171	5202	CLINICAL PSYCHOLOGY	2838.0	568.0	2270.0	
171	172	5203	COUNSELING PSYCHOLOGY	4626.0	931.0	3695.0	
172	173	3501	LIBRARY SCIENCE	1098.0	134.0	964.0	

	Major_category	ShareWomen	Sample_size	Employed	...	\
168	Biology & Life Science	0.637293	47	6259	...	
169	Psychology & Social Work	0.817099	7	2125	...	
170	Psychology & Social Work	0.799859	13	2101	...	
171	Psychology & Social Work	0.798746	21	3777	...	
172	Education	0.877960	2	742	...	

	Part_time	Full_time_year_round	Unemployed	Unemployment_rate	Median	\
168	2190	3602	304	0.046320	26000	
169	572	1211	148	0.065112	25000	
170	648	1293	368	0.149048	25000	
171	965	2738	214	0.053621	23400	
172	237	410	87	0.104946	22000	

	P25th	P75th	College_jobs	Non_college_jobs	Low_wage_jobs
168	20000	39000	2771	2947	743
169	24000	34000	1488	615	82
170	25000	40000	986	870	622
171	19200	26000	2403	1245	308
172	20000	22000	288	338	192

[5 rows x 21 columns]

```
[3]: recent_grads.describe()
```

```
[3]:
```

	Rank	Major_code	Total	Men	Women	\
count	173.000000	173.000000	172.000000	172.000000	172.000000	
mean	87.000000	3879.815029	39370.081395	16723.406977	22646.674419	
std	50.084928	1687.753140	63483.491009	28122.433474	41057.330740	
min	1.000000	1100.000000	124.000000	119.000000	0.000000	
25%	44.000000	2403.000000	4549.750000	2177.500000	1778.250000	
50%	87.000000	3608.000000	15104.000000	5434.000000	8386.500000	
75%	130.000000	5503.000000	38909.750000	14631.000000	22553.750000	
max	173.000000	6403.000000	393735.000000	173809.000000	307087.000000	

	ShareWomen	Sample_size	Employed	Full_time	Part_time	\
count	172.000000	173.000000	173.000000	173.000000	173.000000	
mean	0.522223	356.080925	31192.763006	26029.306358	8832.398844	

std	0.231205	618.361022	50675.002241	42869.655092	14648.179473
min	0.000000	2.000000	0.000000	111.000000	0.000000
25%	0.336026	39.000000	3608.000000	3154.000000	1030.000000
50%	0.534024	130.000000	11797.000000	10048.000000	3299.000000
75%	0.703299	338.000000	31433.000000	25147.000000	9948.000000
max	0.968954	4212.000000	307933.000000	251540.000000	115172.000000

	Full_time_year_round	Unemployed	Unemployment_rate	Median	\
count	173.000000	173.000000	173.000000	173.000000	
mean	19694.427746	2416.329480	0.068191	40151.445087	
std	33160.941514	4112.803148	0.030331	11470.181802	
min	111.000000	0.000000	0.000000	22000.000000	
25%	2453.000000	304.000000	0.050306	33000.000000	
50%	7413.000000	893.000000	0.067961	36000.000000	
75%	16891.000000	2393.000000	0.087557	45000.000000	
max	199897.000000	28169.000000	0.177226	110000.000000	

	P25th	P75th	College_jobs	Non_college_jobs	\
count	173.000000	173.000000	173.000000	173.000000	
mean	29501.445087	51494.219653	12322.635838	13284.497110	
std	9166.005235	14906.279740	21299.868863	23789.655363	
min	18500.000000	22000.000000	0.000000	0.000000	
25%	24000.000000	42000.000000	1675.000000	1591.000000	
50%	27000.000000	47000.000000	4390.000000	4595.000000	
75%	33000.000000	60000.000000	14444.000000	11783.000000	
max	95000.000000	125000.000000	151643.000000	148395.000000	

	Low_wage_jobs
count	173.000000
mean	3859.017341
std	6944.998579
min	0.000000
25%	340.000000
50%	1231.000000
75%	3466.000000
max	48207.000000

```
[4]: raw_data_count = recent_grads.shape
raw_data_count
```

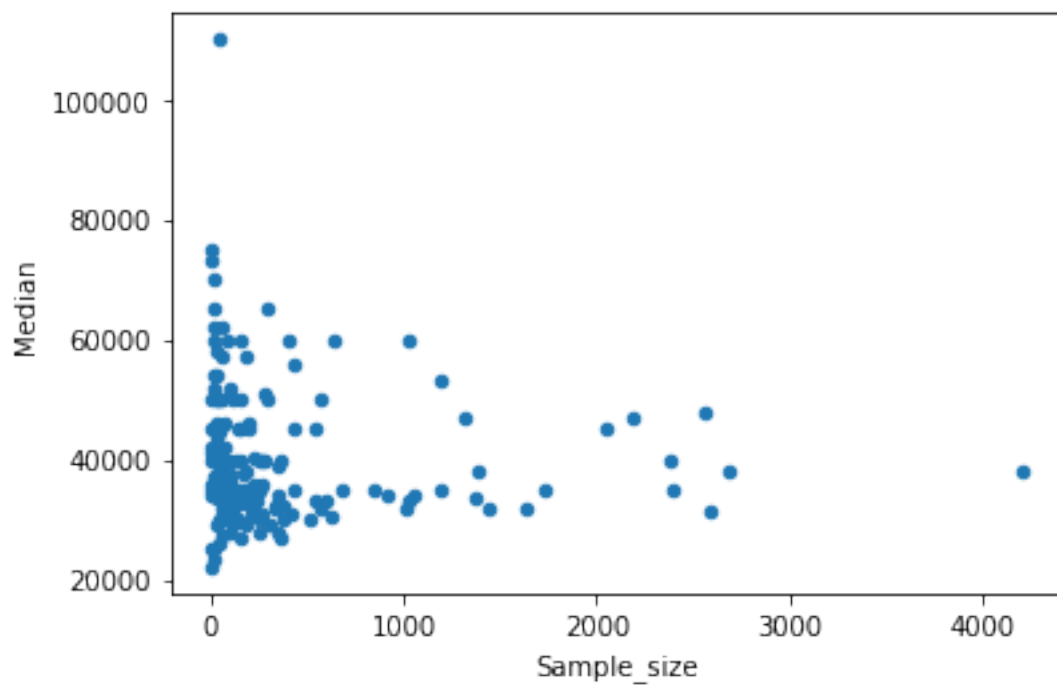
```
[4]: (173, 21)
```

```
[5]: recent_grads = recent_grads.dropna()
cleaned_data_count = recent_grads.shape
cleaned_data_count
```

```
[5]: (172, 21)
```

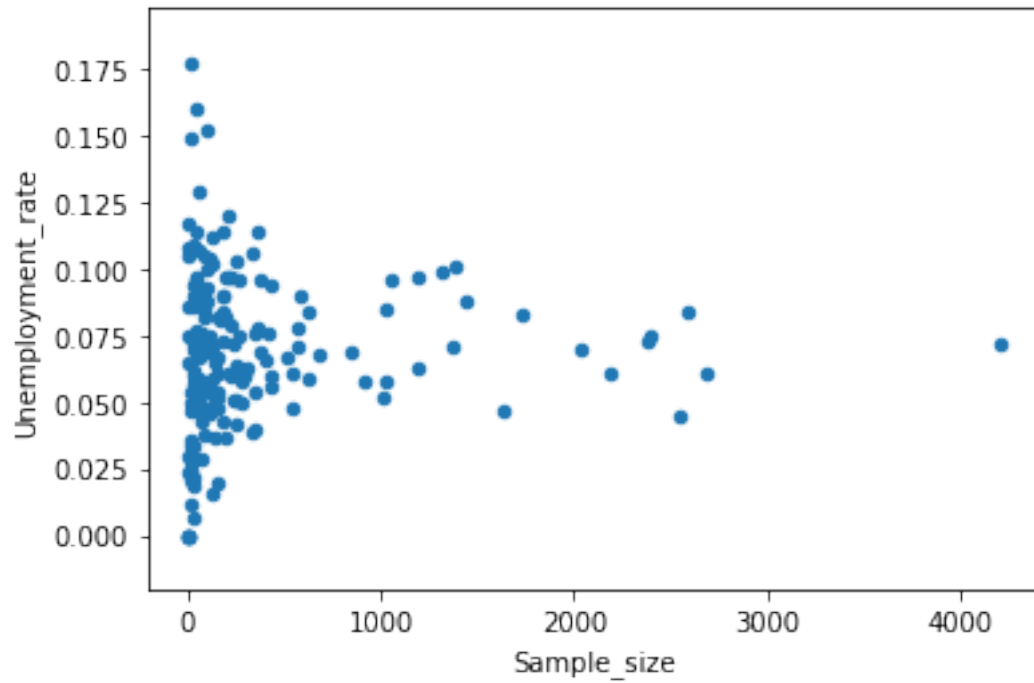
```
[6]: recent_grads.plot(x='Sample_size', y='Median', kind='scatter')
```

```
[6]: <matplotlib.axes._subplots.AxesSubplot at 0x117221ed0>
```



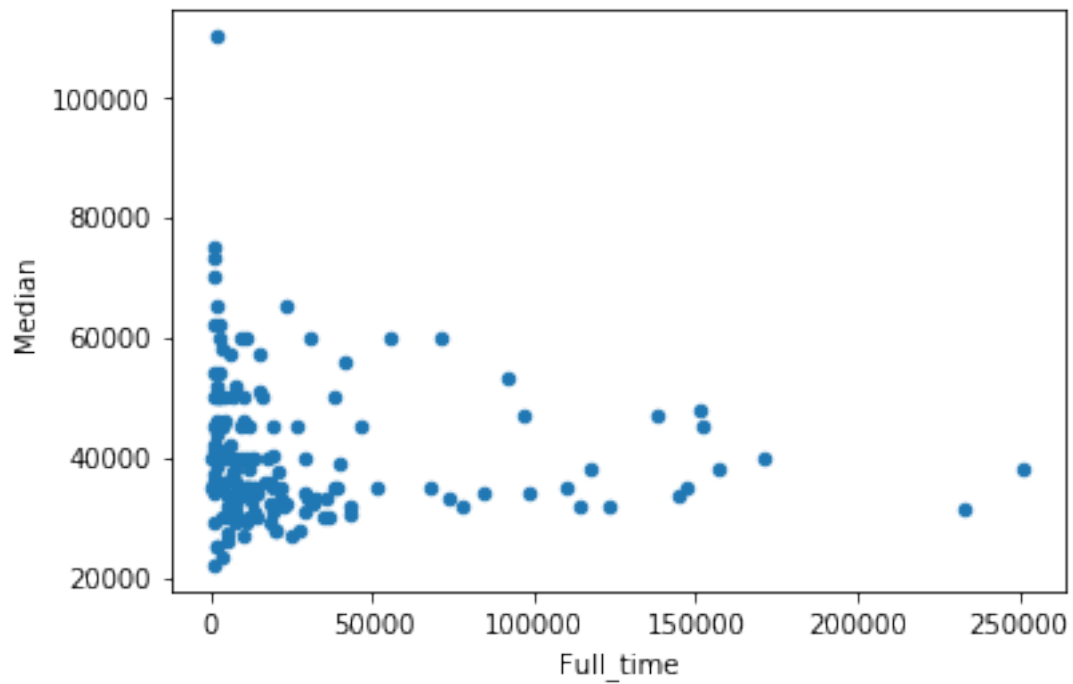
```
[7]: recent_grads.plot(x='Sample_size', y='Unemployment_rate', kind='scatter')
```

```
[7]: <matplotlib.axes._subplots.AxesSubplot at 0x117d0f790>
```



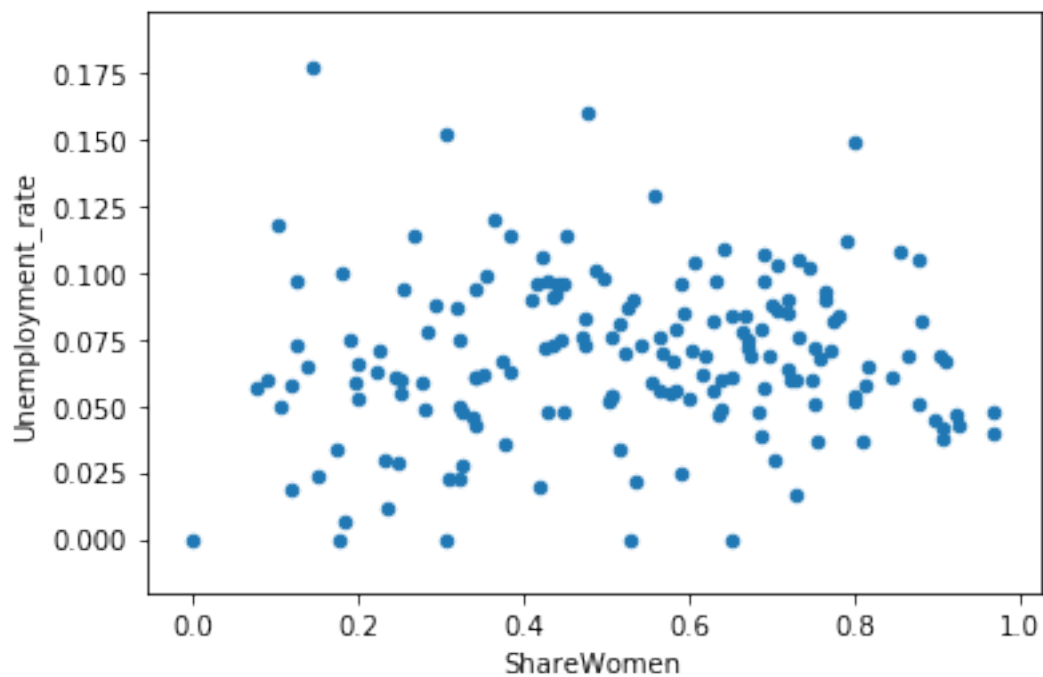
```
[8]: recent_grads.plot(x='Full_time', y='Median', kind='scatter')
```

```
[8]: <matplotlib.axes._subplots.AxesSubplot at 0x117e232d0>
```



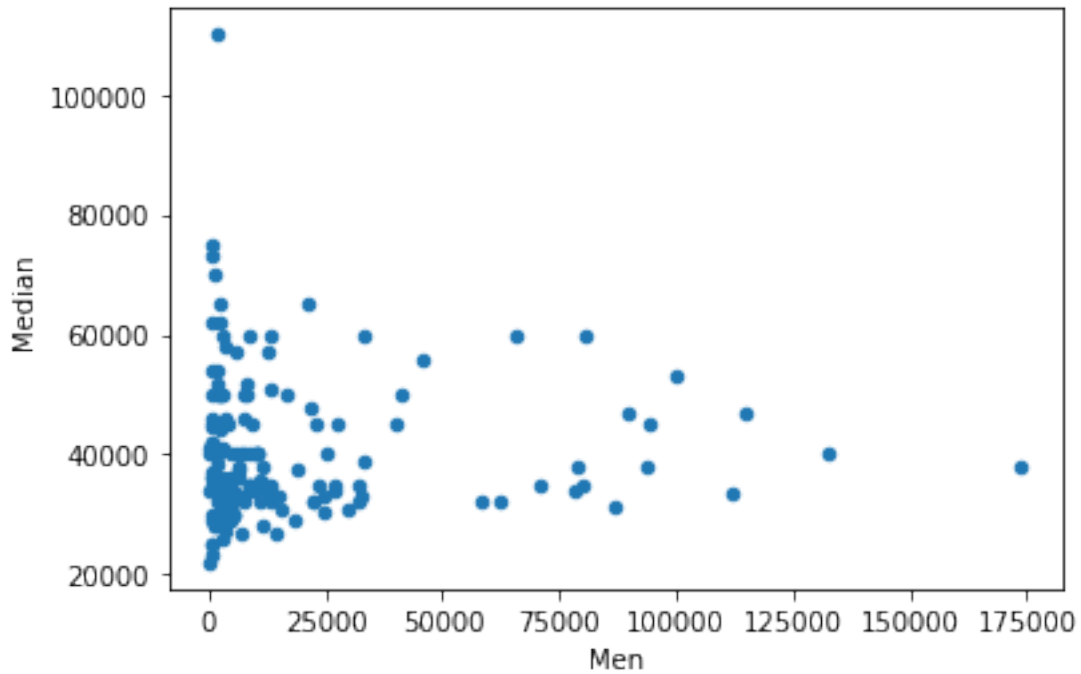
```
[9]: recent_grads.plot(x='ShareWomen', y='Unemployment_rate', kind='scatter')
```

```
[9]: <matplotlib.axes._subplots.AxesSubplot at 0x117f34410>
```



```
[10]: recent_grads.plot(x='Men', y='Median', kind='scatter')
```

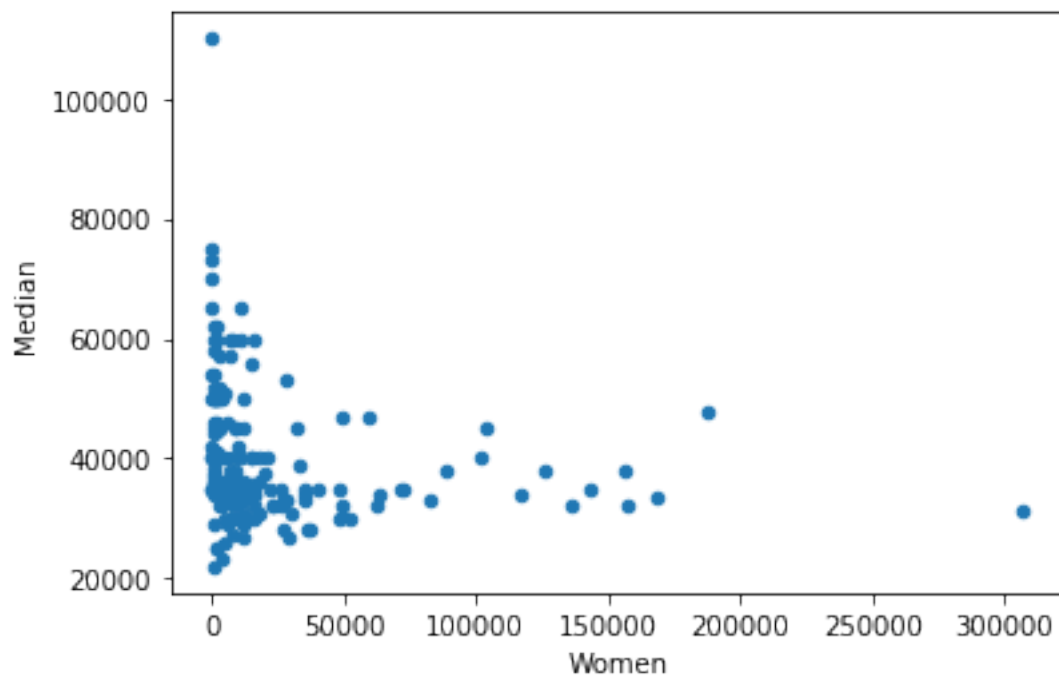
```
[10]: <matplotlib.axes._subplots.AxesSubplot at 0x11801be10>
```



Dependent variables: values on the y-axis Independent variables: values on x-axis - Things to lookk at in a scatter plot: Direction - positive or negative, Form - linear or non-linear, Strength - weak or moderate or strong, outliers

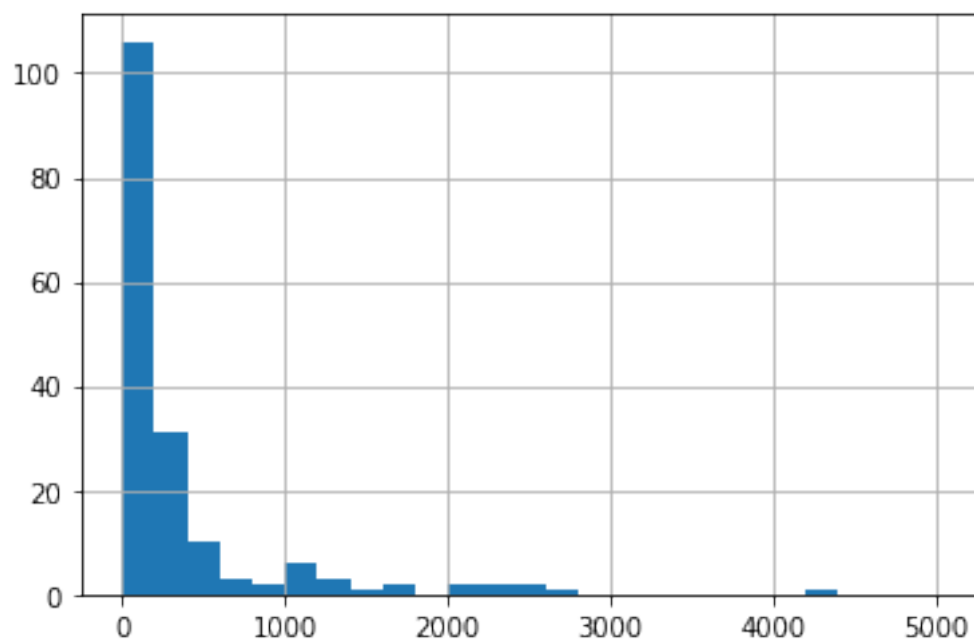
```
[11]: recent_grads.plot(x='Women', y='Median', kind='scatter')
```

```
[11]: <matplotlib.axes._subplots.AxesSubplot at 0x1180e7850>
```



```
[12]: recent_grads['Sample_size'].hist(bins=25, range=(0,5000))
```

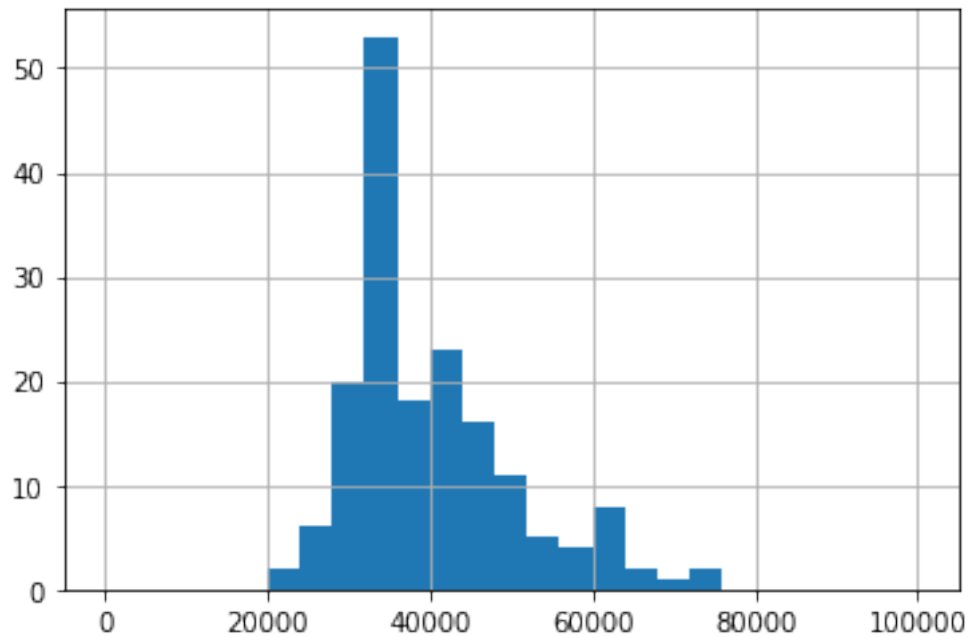
```
[12]: <matplotlib.axes._subplots.AxesSubplot at 0x1181d6a10>
```





```
[13]: recent_grads['Median'].hist(bins=25, range=(0,100000))
```

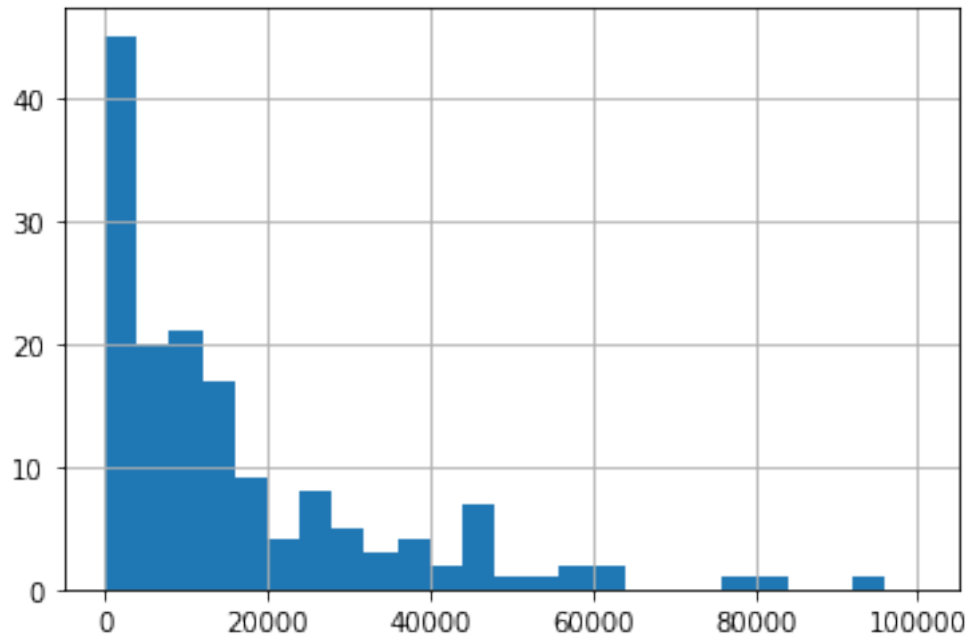
```
[13]: <matplotlib.axes._subplots.AxesSubplot at 0x1171f1590>
```



As we can see from the graph above, the most common median salary ranges anywhere between 30,000 and 35,000 dollars.

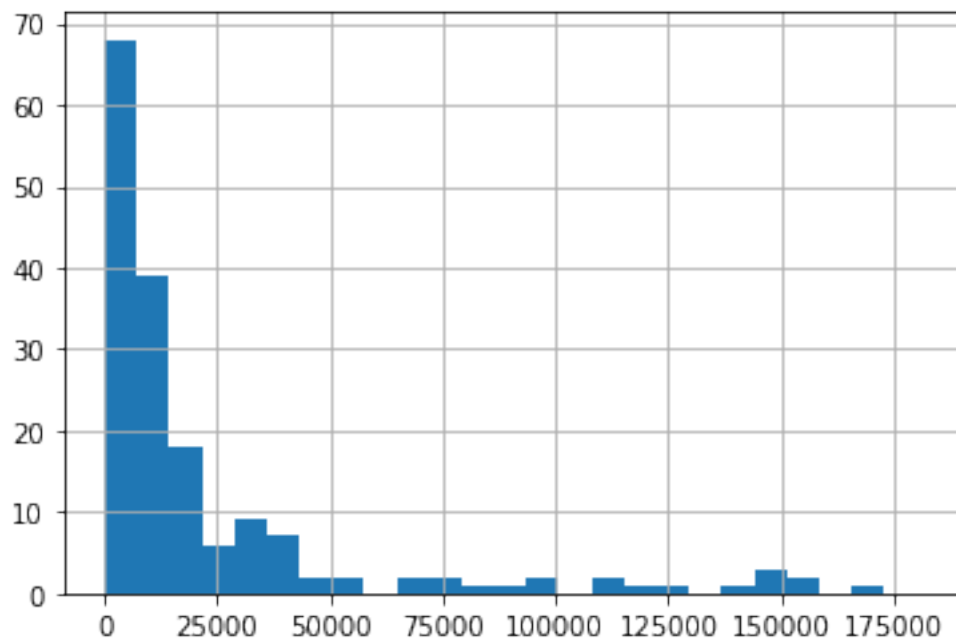
```
[14]: recent_grads['Employed'].hist(bins=25, range=(0,100000))
```

```
[14]: <matplotlib.axes._subplots.AxesSubplot at 0x11842f510>
```



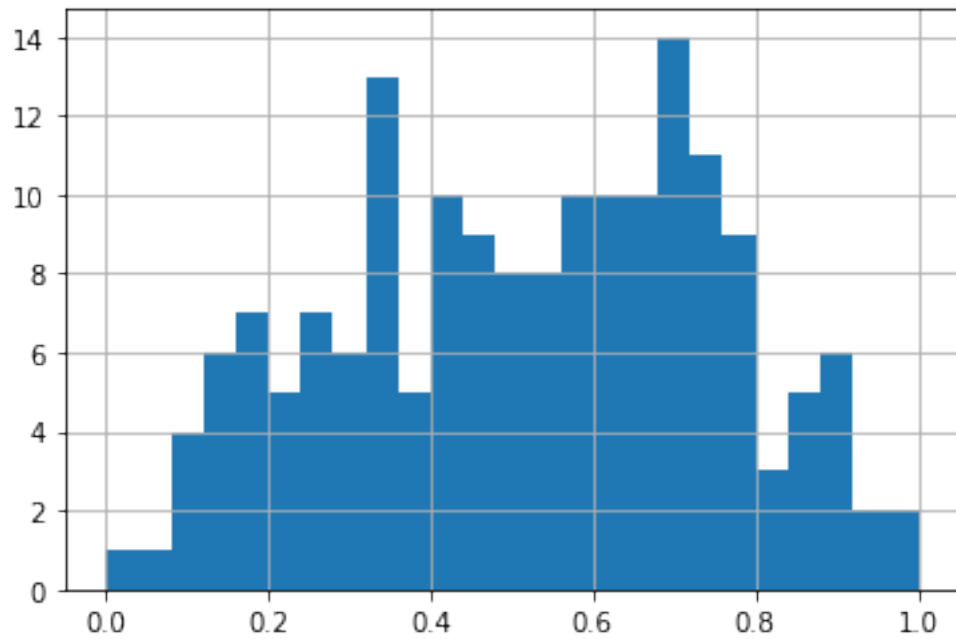
```
[15]: recent_grads['Full_time'].hist(bins=25, range=(0,180000))
```

```
[15]: <matplotlib.axes._subplots.AxesSubplot at 0x118549410>
```



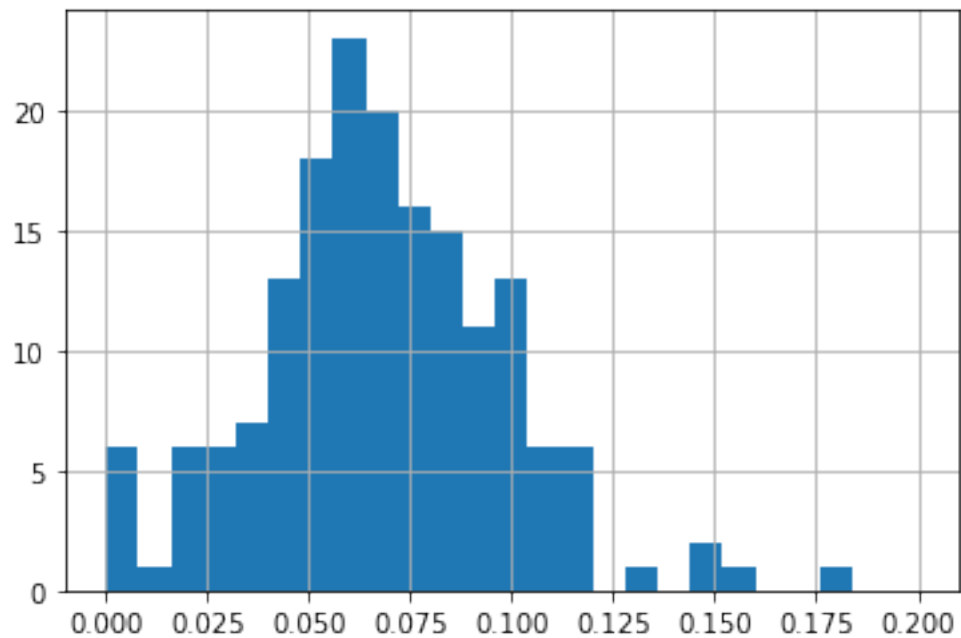
```
[16]: recent_grads['ShareWomen'].hist(bins=25, range=(0,1))
```

```
[16]: <matplotlib.axes._subplots.AxesSubplot at 0x118664c10>
```



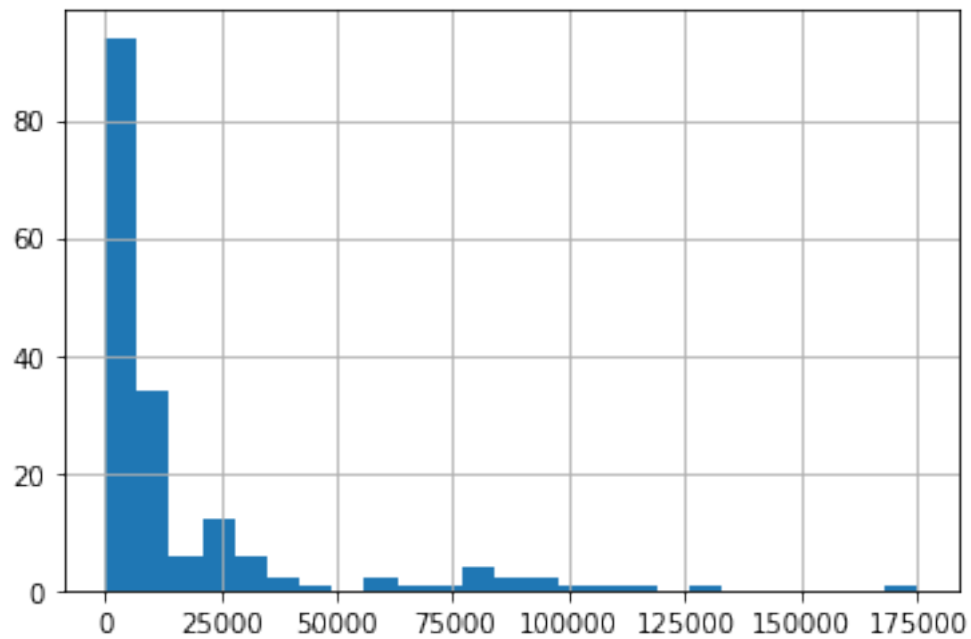
```
[17]: recent_grads['Unemployment_rate'].hist(bins=25, range=(0,0.2))
```

```
[17]: <matplotlib.axes._subplots.AxesSubplot at 0x11878f7d0>
```



```
[18]: recent_grads['Men'].hist(bins=25, range=(0,175000))
```

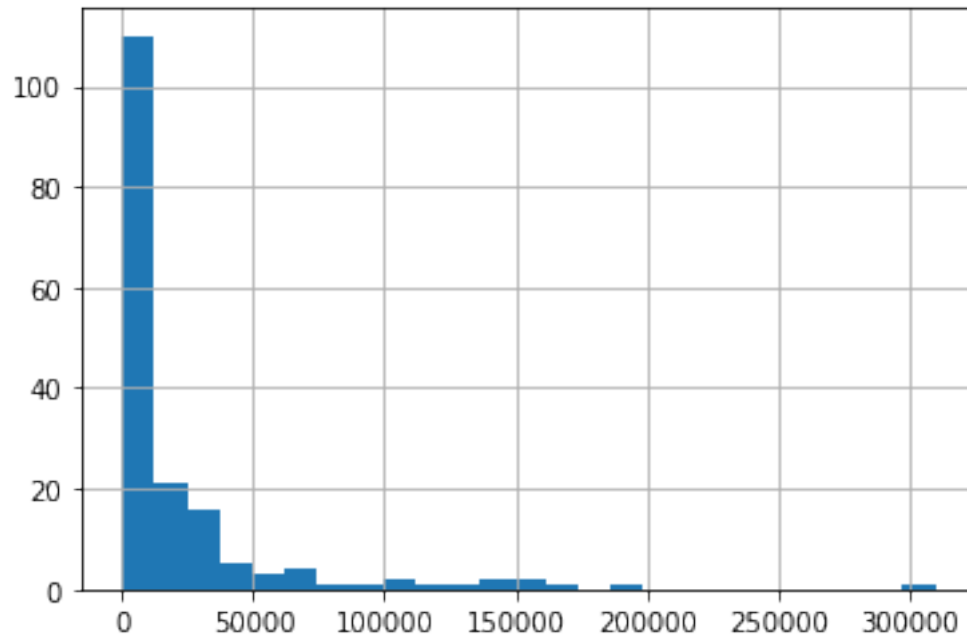
```
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x1188a3e90>
```



As you can observe from the graph above, about 55% of the majors were predominantly men.

```
[19]: recent_grads['Women'].hist(bins=25, range=(0,310000))
```

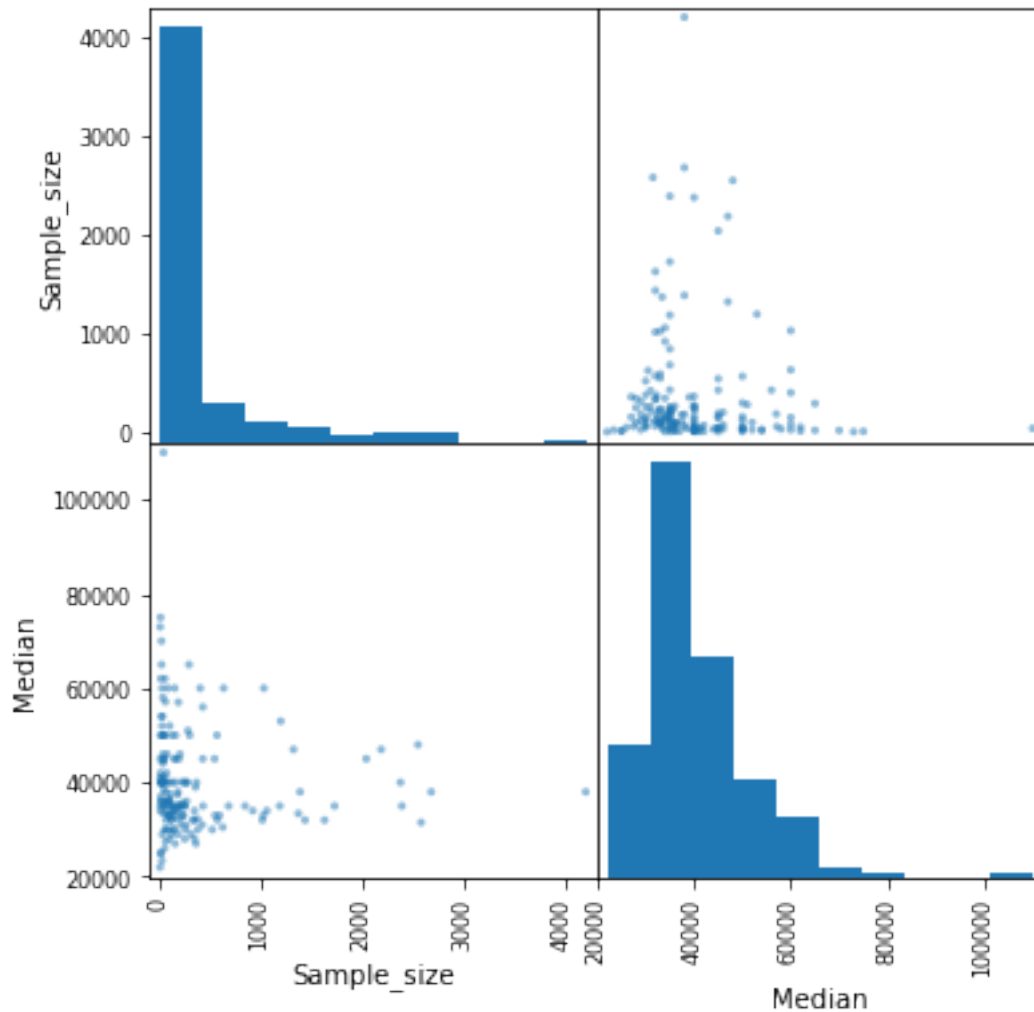
```
[19]: <matplotlib.axes._subplots.AxesSubplot at 0x1189d0490>
```



As you can observe from the graph above, about 66-70% of majors were predominantly women.

```
[26]: from pandas.plotting import scatter_matrix
      scatter_matrix(recent_grads[['Sample_size', 'Median']], figsize=(6,6))
```

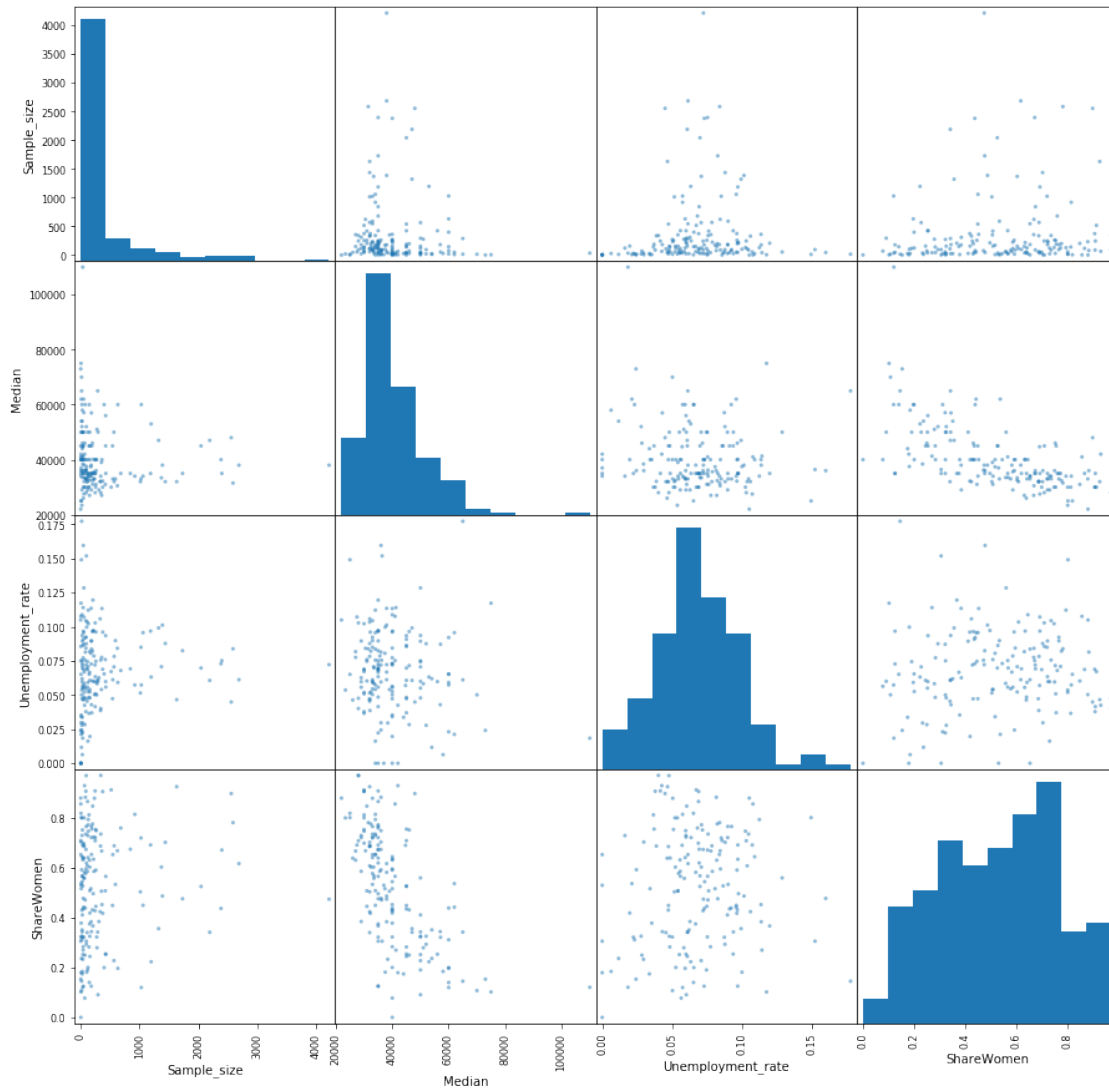
```
[26]: array([[<matplotlib.axes._subplots.AxesSubplot object at 0x118f1e890>,
               <matplotlib.axes._subplots.AxesSubplot object at 0x118fd0710>],
               [<matplotlib.axes._subplots.AxesSubplot object at 0x1190b7f10>,
               <matplotlib.axes._subplots.AxesSubplot object at 0x1190ec710>]],
               dtype=object)
```



```
[30]: scatter_matrix(recent_grads[['Sample_size', 'Median', 'Unemployment_rate', 'ShareWomen']], figsize=(15,15))
```

```
[30]: array([[<matplotlib.axes._subplots.AxesSubplot object at 0x11a5d7390>,
<matplotlib.axes._subplots.AxesSubplot object at 0x104ec5610>,
<matplotlib.axes._subplots.AxesSubplot object at 0x104ef0dd0>,
<matplotlib.axes._subplots.AxesSubplot object at 0x104f31610>],
[<matplotlib.axes._subplots.AxesSubplot object at 0x105c4be10>,
<matplotlib.axes._subplots.AxesSubplot object at 0x105c8c650>,
<matplotlib.axes._subplots.AxesSubplot object at 0x105cc0e50>,
<matplotlib.axes._subplots.AxesSubplot object at 0x105d01690>],
[<matplotlib.axes._subplots.AxesSubplot object at 0x105d0b6d0>,
<matplotlib.axes._subplots.AxesSubplot object at 0x105d4d090>,
<matplotlib.axes._subplots.AxesSubplot object at 0x119a45f10>,
<matplotlib.axes._subplots.AxesSubplot object at 0x119a7b710>],
```

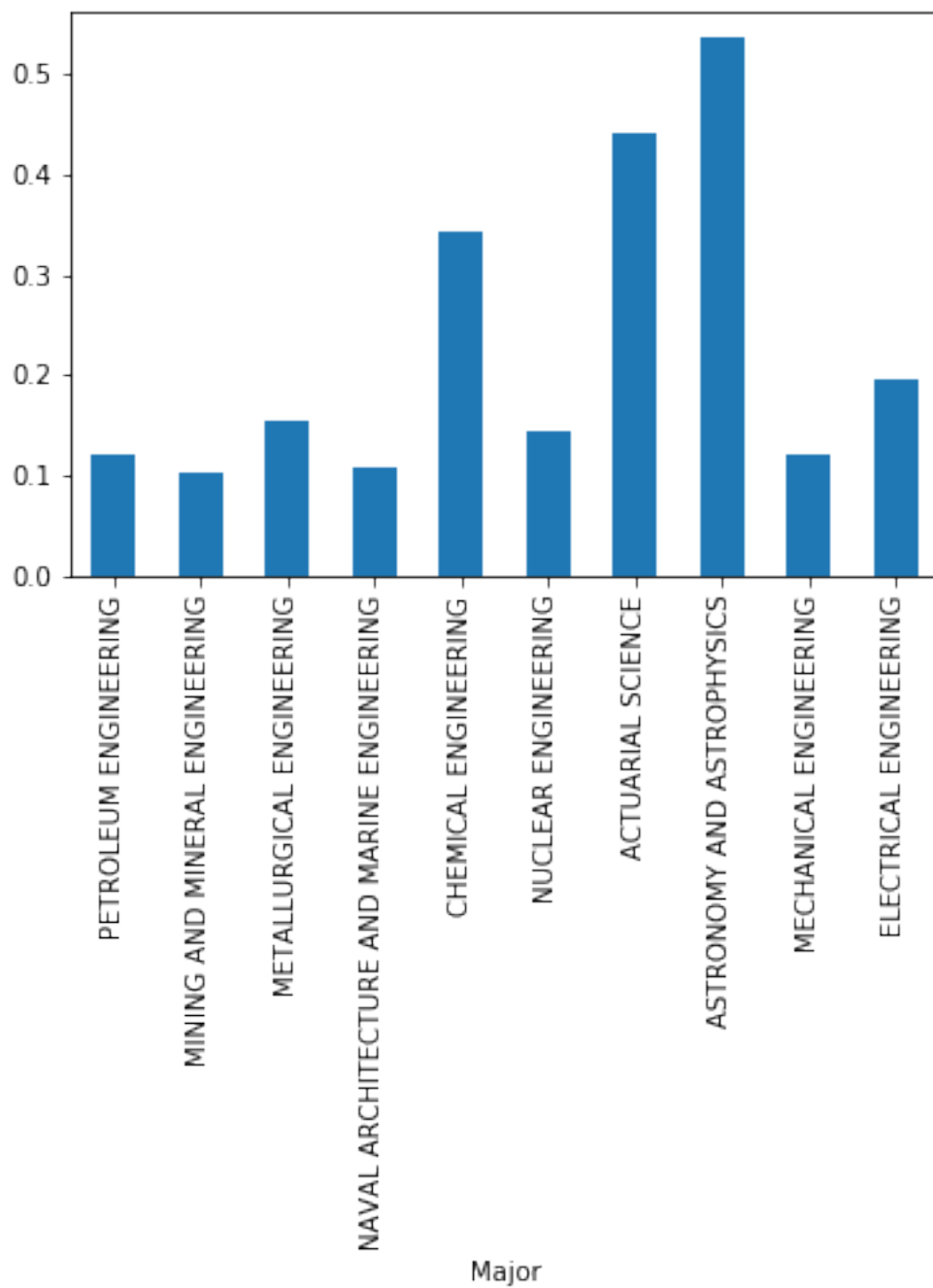
```
[<matplotlib.axes._subplots.AxesSubplot object at 0x119abaf50>,
 <matplotlib.axes._subplots.AxesSubplot object at 0x119af0750>,
 <matplotlib.axes._subplots.AxesSubplot object at 0x119b24f50>,
 <matplotlib.axes._subplots.AxesSubplot object at 0x119b67790>]],
 dtype=object)
```



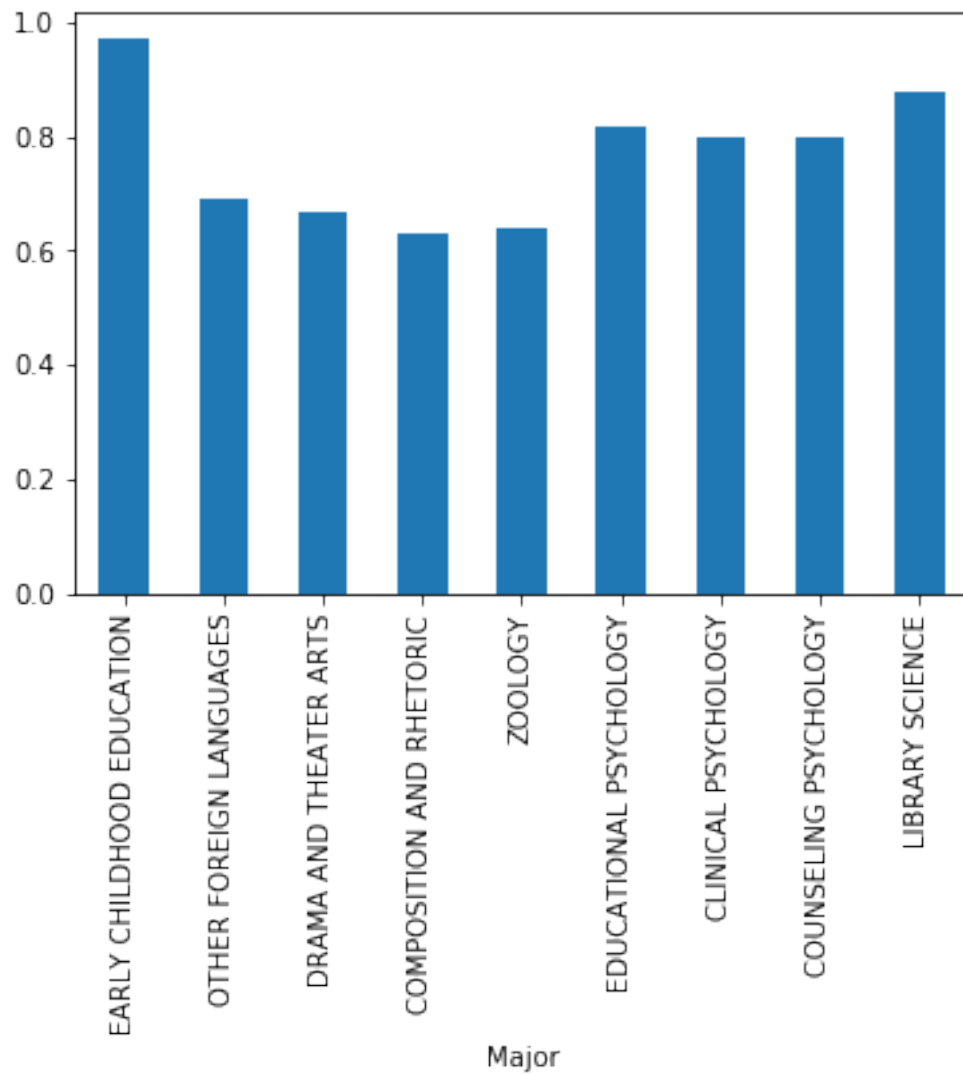
- In a sample size of about 500 people, most earn a Median salary between 20,000 to 40,000 dollars.
- In the majors that were majority women, those majors earned significantly less than majors that were majority men.
- In the first sample size of about 1000 people, most of them were women.

```
[39]: recent_grads[:10].plot.bar(x='Major', y='ShareWomen', legend=False)
recent_grads[163:].plot.bar(x='Major', y='ShareWomen', legend=False)
```

[39]: <matplotlib.axes.\_subplots.AxesSubplot at 0x11b5f3610>

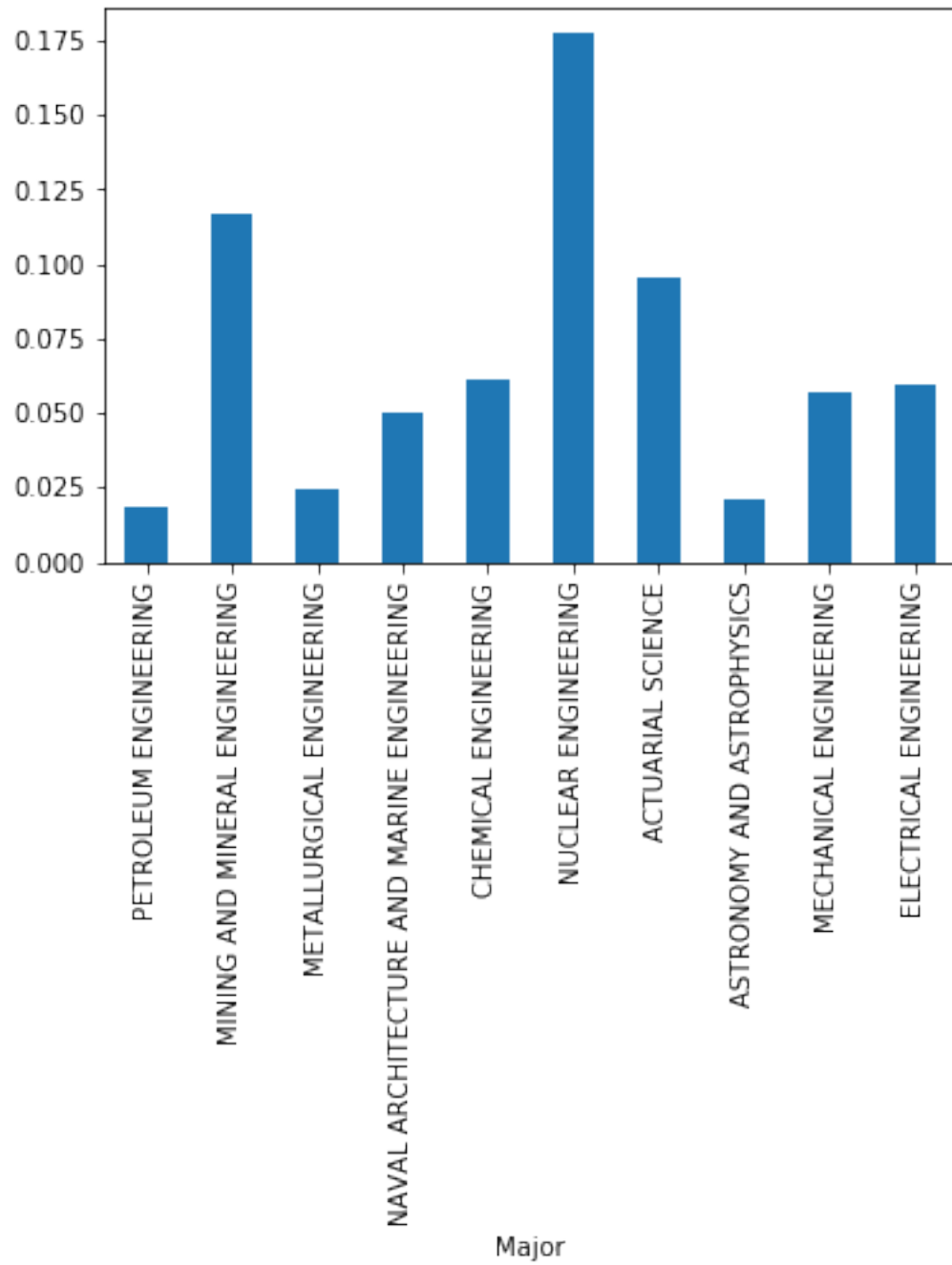


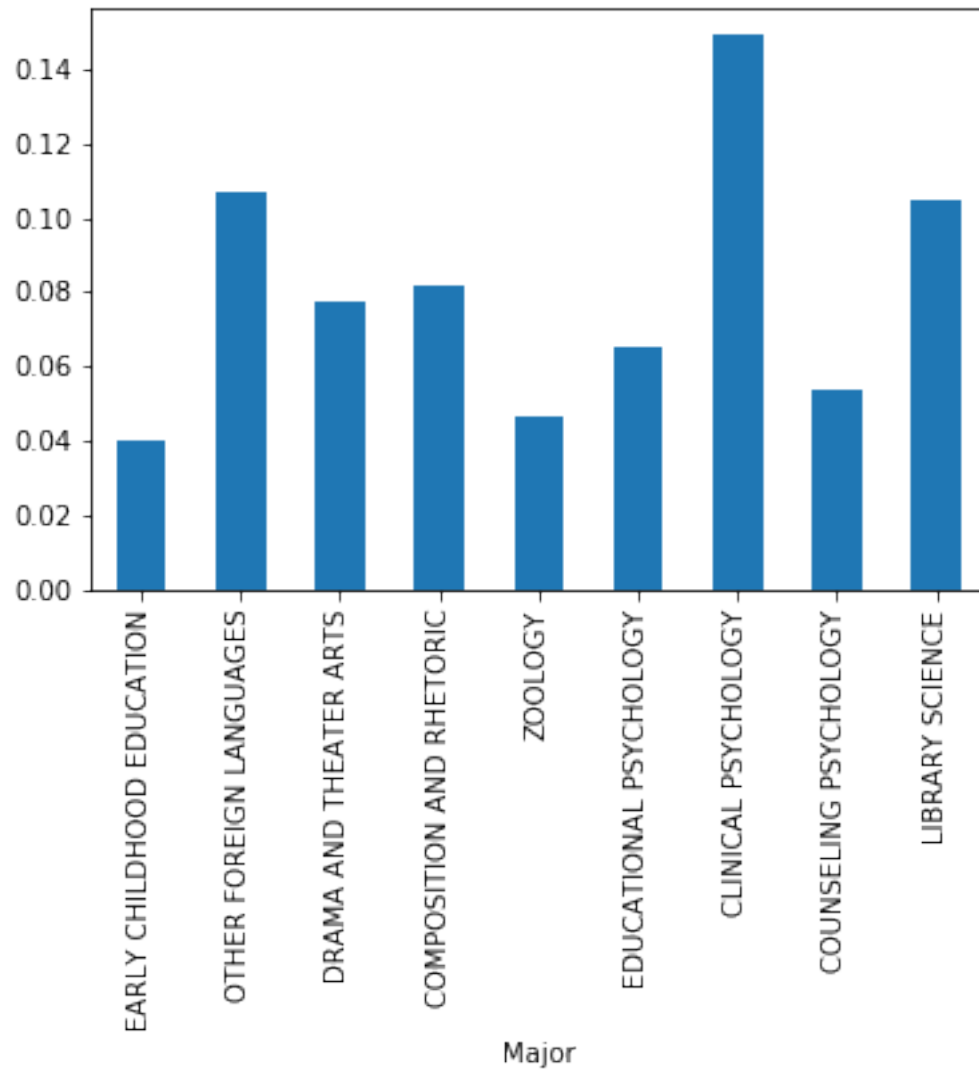




```
[41]: recent_grads[:10].plot.bar(x='Major', y='Unemployment_rate', legend=False)
recent_grads[163:].plot.bar(x='Major', y='Unemployment_rate', legend=False)
```

```
[41]: <matplotlib.axes._subplots.AxesSubplot at 0x11b7d5a90>
```





[ ]: