

seminarioCD4_graph_pandas

August 27, 2025

```
[6]: import pandas as pd
```

```
[7]: url="/Users/marcobarragan/Documents/Seminario-de-Ciencia-de-datos/salary.csv"
data = pd.read_csv(url)
```

```
[8]: data.head()
```

```
[8]:
```

	age	workclass	fnlwgt	education	education-num	\
0	39	State-gov	77516	Bachelors	13	
1	50	Self-emp-not-inc	83311	Bachelors	13	
2	38	Private	215646	HS-grad	9	
3	53	Private	234721	11th	7	
4	28	Private	338409	Bachelors	13	

	marital-status	occupation	relationship	race	sex	\
0	Never-married	Adm-clerical	Not-in-family	White	Male	
1	Married-civ-spouse	Exec-managerial	Husband	White	Male	
2	Divorced	Handlers-cleaners	Not-in-family	White	Male	
3	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male	
4	Married-civ-spouse	Prof-specialty	Wife	Black	Female	

	capital-gain	capital-loss	hours-per-week	native-country	salary
0	2174	0	40	United-States	<=50K
1	0	0	13	United-States	<=50K
2	0	0	40	United-States	<=50K
3	0	0	40	United-States	<=50K
4	0	0	40	Cuba	<=50K

```
[9]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32561 entries, 0 to 32560
Data columns (total 15 columns):
#   Column          Non-Null Count  Dtype
---  -
0   age             32561 non-null  int64
1   workclass       32561 non-null  object
2   fnlwgt          32561 non-null  int64
```

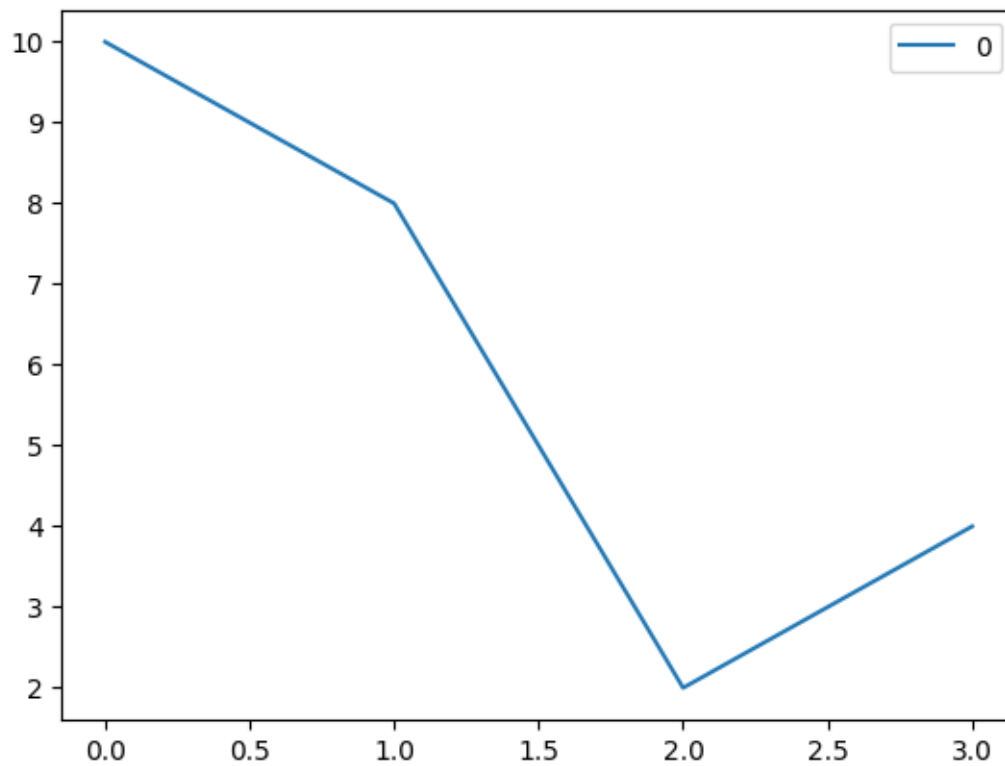
```

3  education      32561 non-null  object
4  education-num  32561 non-null  int64
5  marital-status 32561 non-null  object
6  occupation     32561 non-null  object
7  relationship   32561 non-null  object
8  race           32561 non-null  object
9  sex            32561 non-null  object
10 capital-gain   32561 non-null  int64
11 capital-loss   32561 non-null  int64
12 hours-per-week 32561 non-null  int64
13 native-country 32561 non-null  object
14 salary         32561 non-null  object
dtypes: int64(6), object(9)
memory usage: 3.7+ MB

```

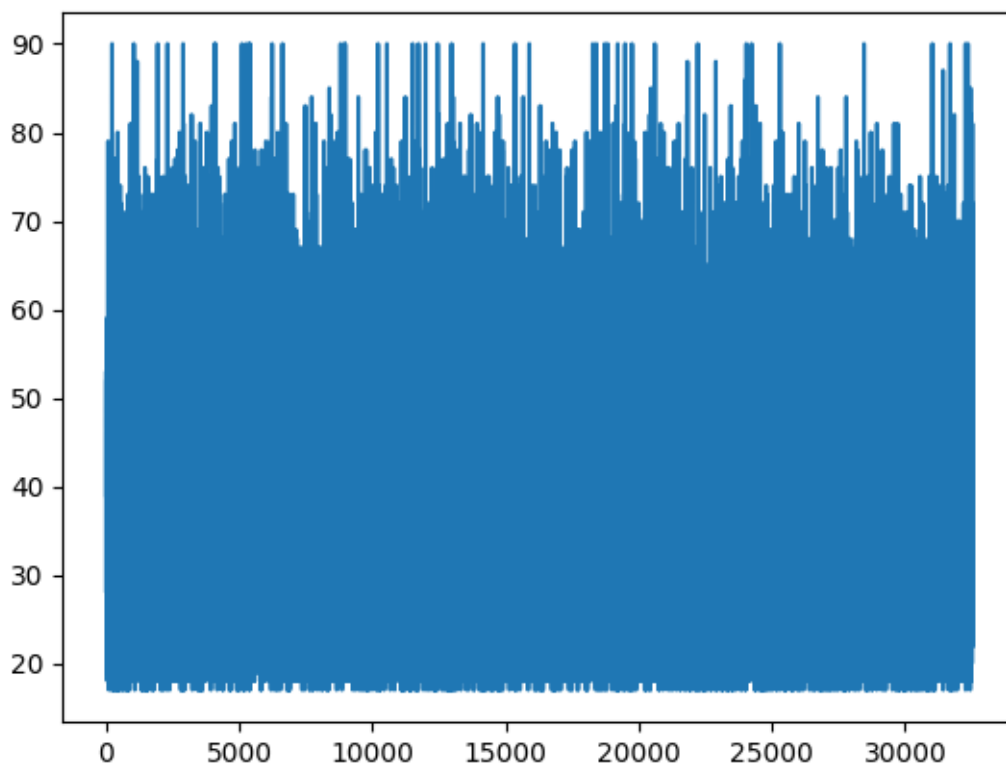
```
[10]: datos=pd.DataFrame([10,8,2,4])
      datos.plot()
```

[10]: <Axes: >



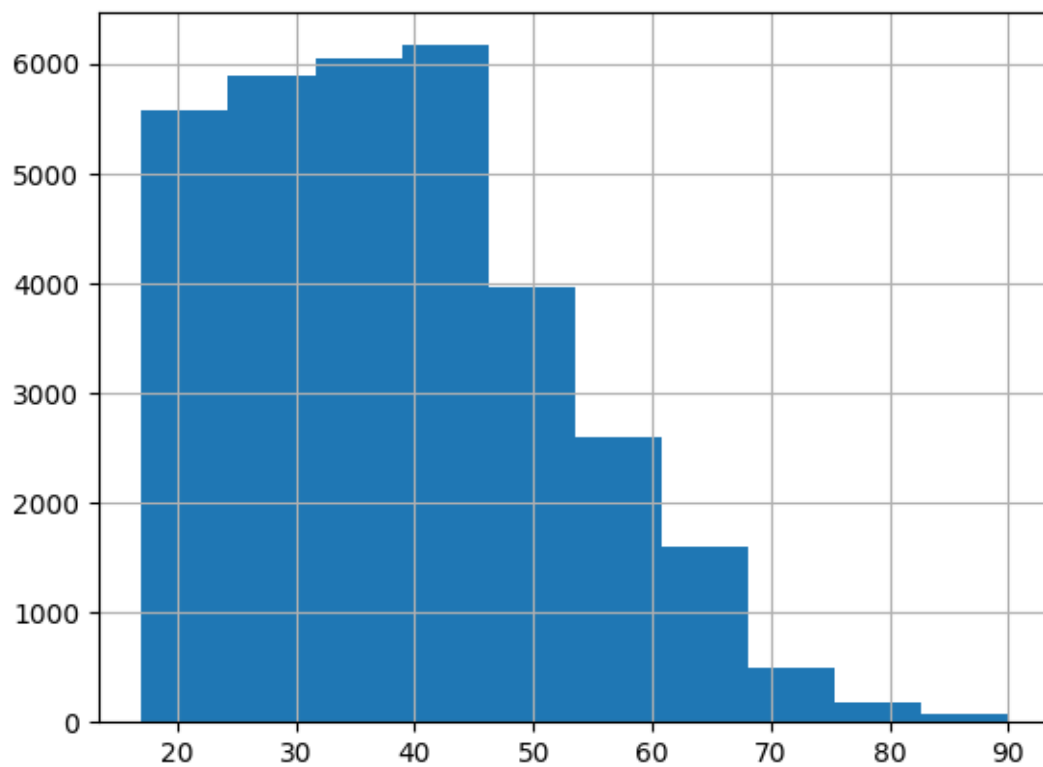
```
[11]: data["age"].plot()
```

[11]: <Axes: >



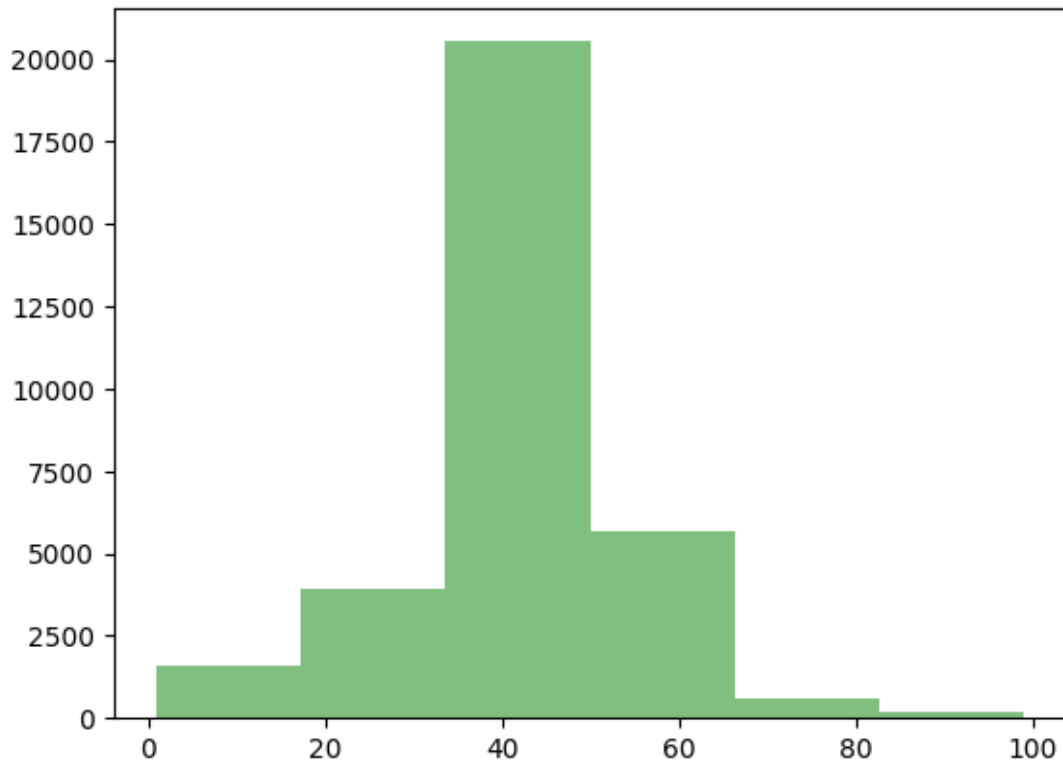
```
[12]: data["age"].hist()
```

[12]: <Axes: >



```
[13]: data["hours-per-week"].hist(bins=6, grid=False, color="Green", alpha=0.5)
```

```
[13]: <Axes: >
```



0.1 Box plot

```
[14]: from scipy import stats
```

```
[15]: p75ages = stats.scoreatpercentile(data["age"], 75)
print("Percentil 75 = {}".format(p75ages))

p10ages = stats.scoreatpercentile(data["age"], 10)

print("Percentil 10 = {}".format(p10ages))

q2ages = stats.scoreatpercentile(data["age"], 50)

print("Cuartil 2 = {}".format(q2ages))
```

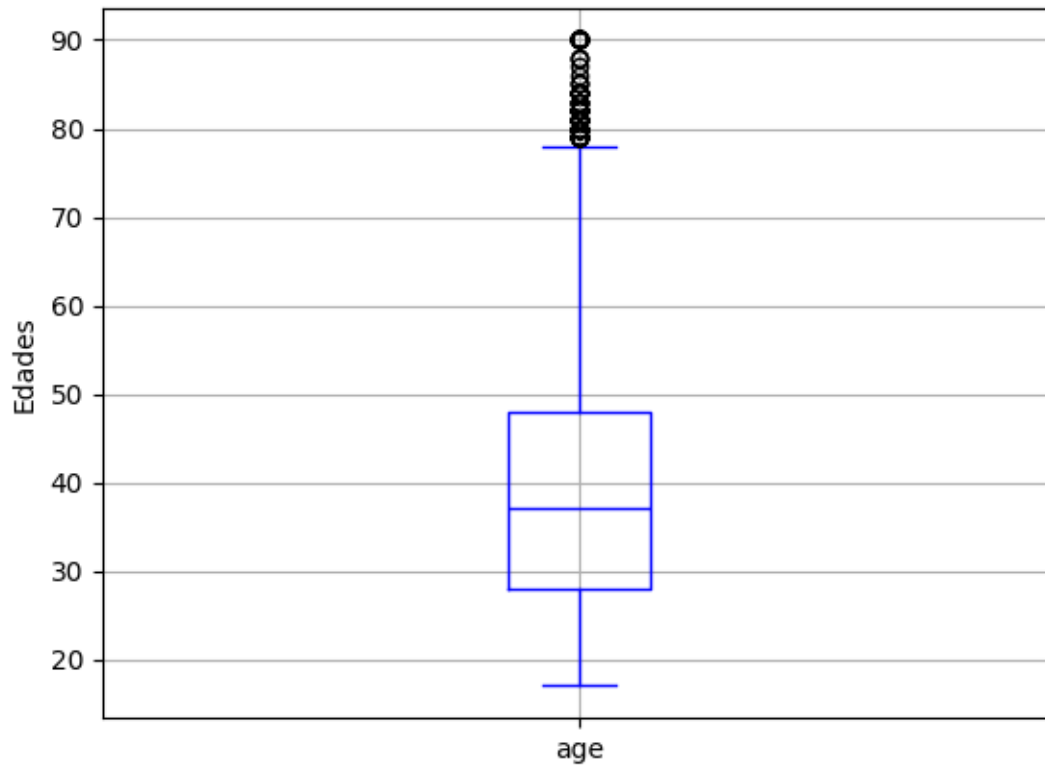
Percentil 75 = 48.0

Percentil 10 = 22.0

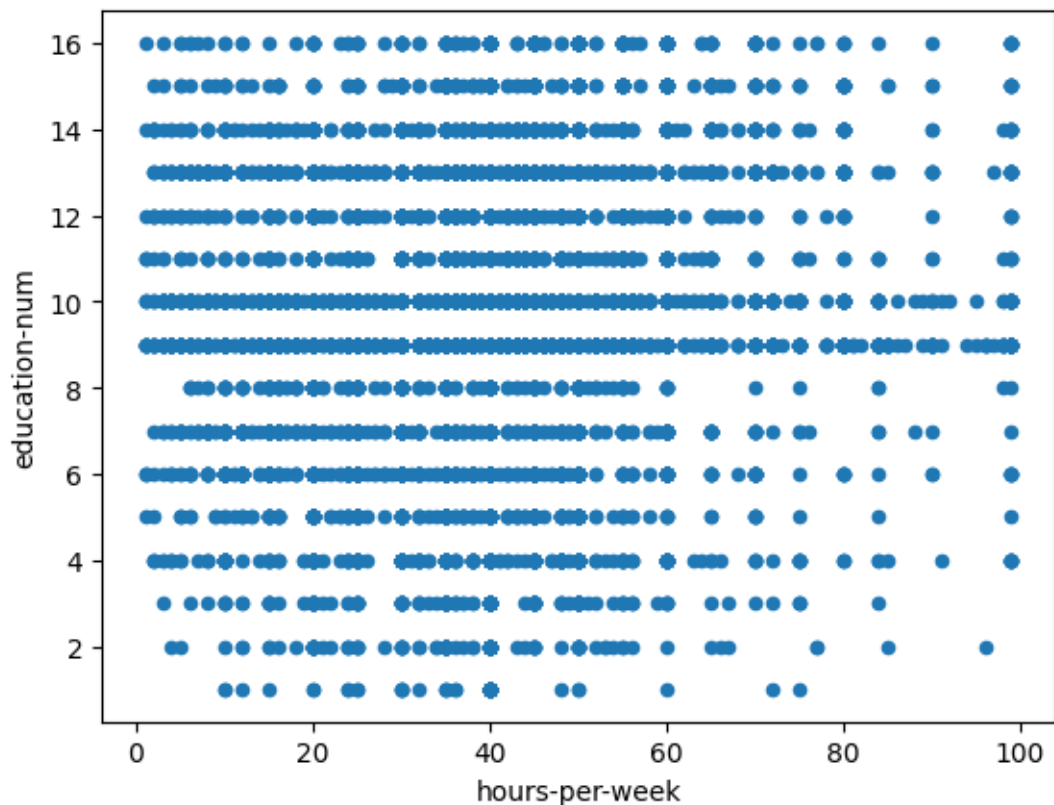
Cuartil 2 = 37.0

```
[36]: data["age"].plot.box(grid=True, ylabel="Edades", color="blue")
```

```
[36]: <Axes: ylabel='Edades'>
```



```
[37]: data.plot.scatter(x='hours-per-week', y='education-num');
```



```
[18]: data.head()
```

```
[18]:
```

	age	workclass	fnlwgt	education	education-num	\
0	39	State-gov	77516	Bachelors	13	
1	50	Self-emp-not-inc	83311	Bachelors	13	
2	38	Private	215646	HS-grad	9	
3	53	Private	234721	11th	7	
4	28	Private	338409	Bachelors	13	

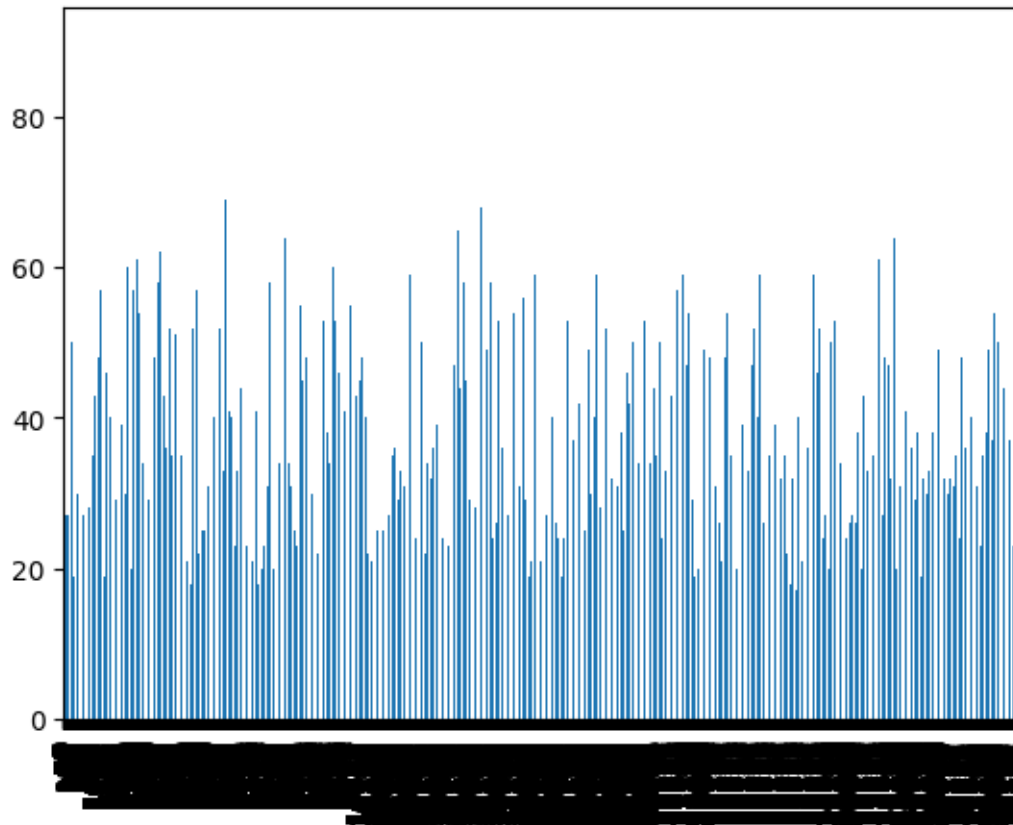
	marital-status	occupation	relationship	race	sex	\
0	Never-married	Adm-clerical	Not-in-family	White	Male	
1	Married-civ-spouse	Exec-managerial	Husband	White	Male	
2	Divorced	Handlers-cleaners	Not-in-family	White	Male	
3	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male	
4	Married-civ-spouse	Prof-specialty	Wife	Black	Female	

	capital-gain	capital-loss	hours-per-week	native-country	salary
0	2174	0	40	United-States	<=50K
1	0	0	13	United-States	<=50K
2	0	0	40	United-States	<=50K
3	0	0	40	United-States	<=50K

4 0 0 40 Cuba <=50K

```
[19]: data["age"].plot.bar()
```

```
[19]: <Axes: >
```



```
[20]: tab_fr_edad=data['age'].value_counts()
```

```
[21]: tab_fr_edad
```

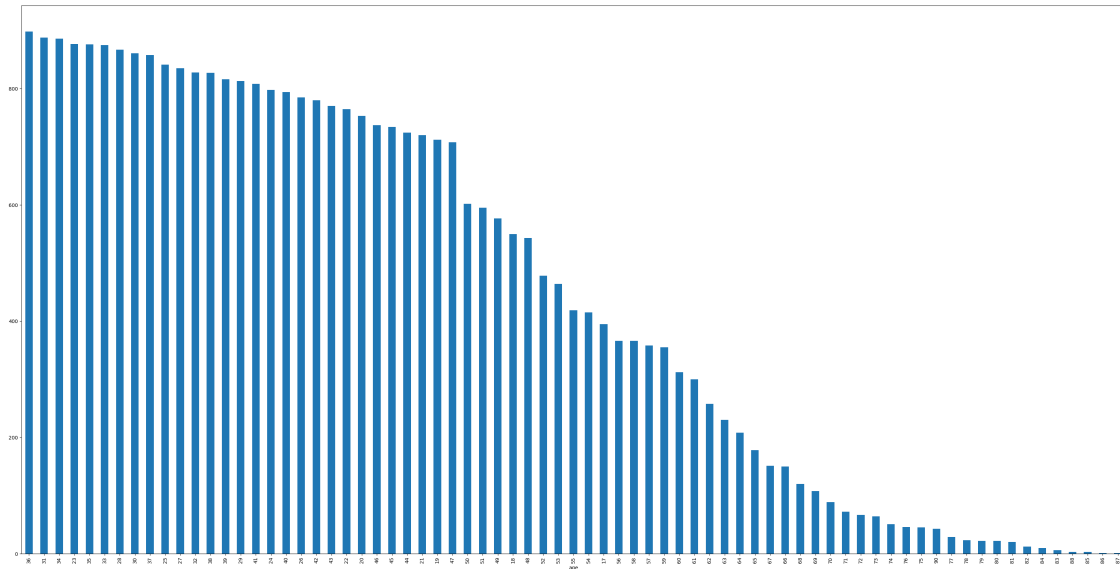
```
[21]: age
36    898
31    888
34    886
23    877
35    876
...
83     6
88     3
85     3
```



```
86      1
87      1
Name: count, Length: 73, dtype: int64
```

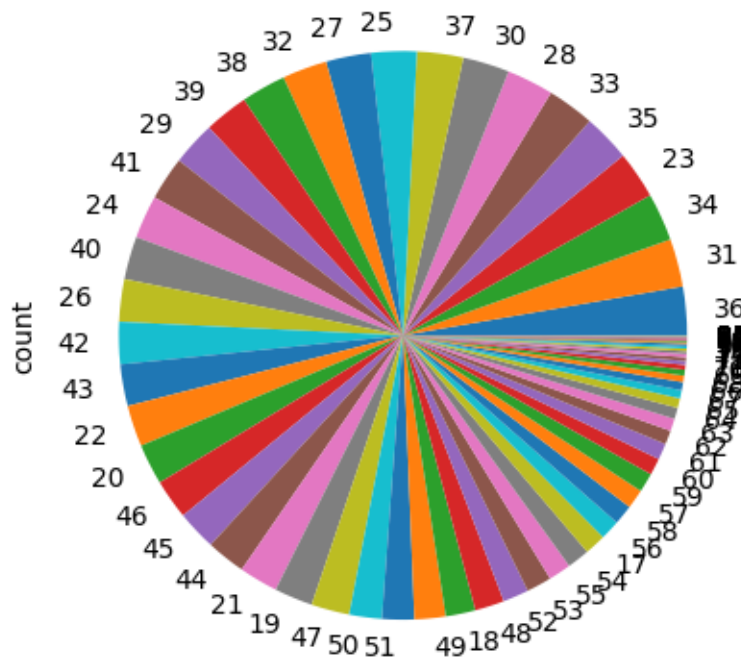
```
[22]: tab_fr_edad.plot.bar(figsize=(40,20))
```

```
[22]: <Axes: xlabel='age'>
```



```
[23]: tab_fr_edad.plot.pie()
```

```
[23]: <Axes: ylabel='count'>
```

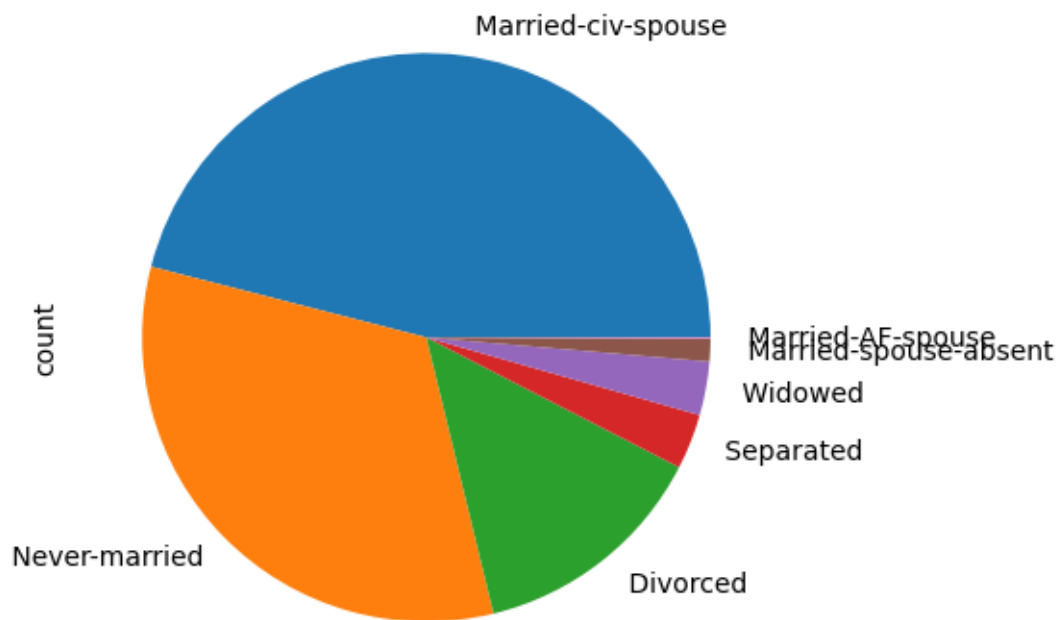


```
[24]: tab_fr_mar_est=data['marital-status'].value_counts()
      tab_fr_mar_est
```

```
[24]: marital-status
      Married-civ-spouse      14976
      Never-married         10683
      Divorced               4443
      Separated              1025
      Widowed                993
      Married-spouse-absent   418
      Married-AF-spouse       23
      Name: count, dtype: int64
```

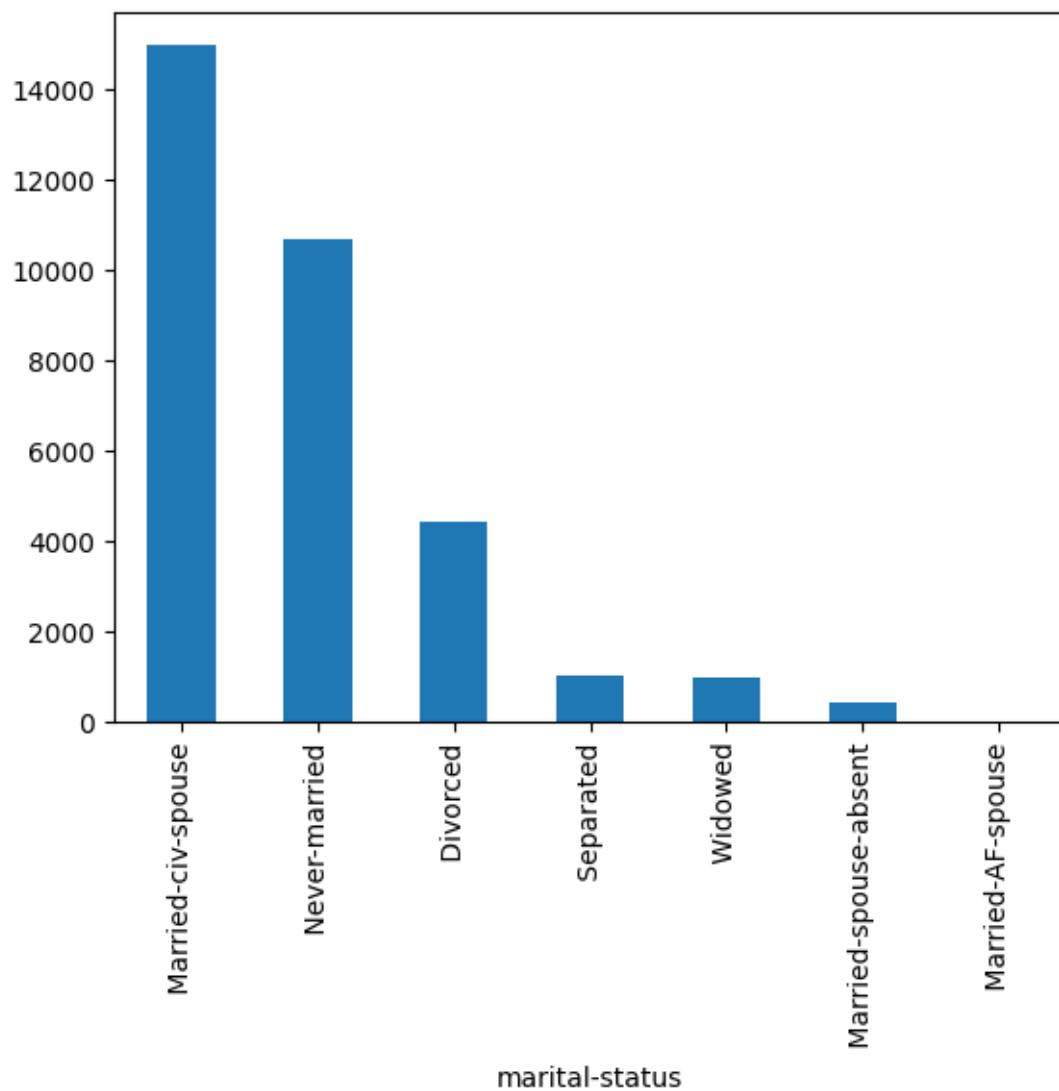
```
[25]: tab_fr_mar_est.plot.pie()
```

```
[25]: <Axes: ylabel='count'>
```



```
[26]: tab_fr_mar_est.plot.bar()
```

```
[26]: <Axes: xlabel='marital-status'>
```



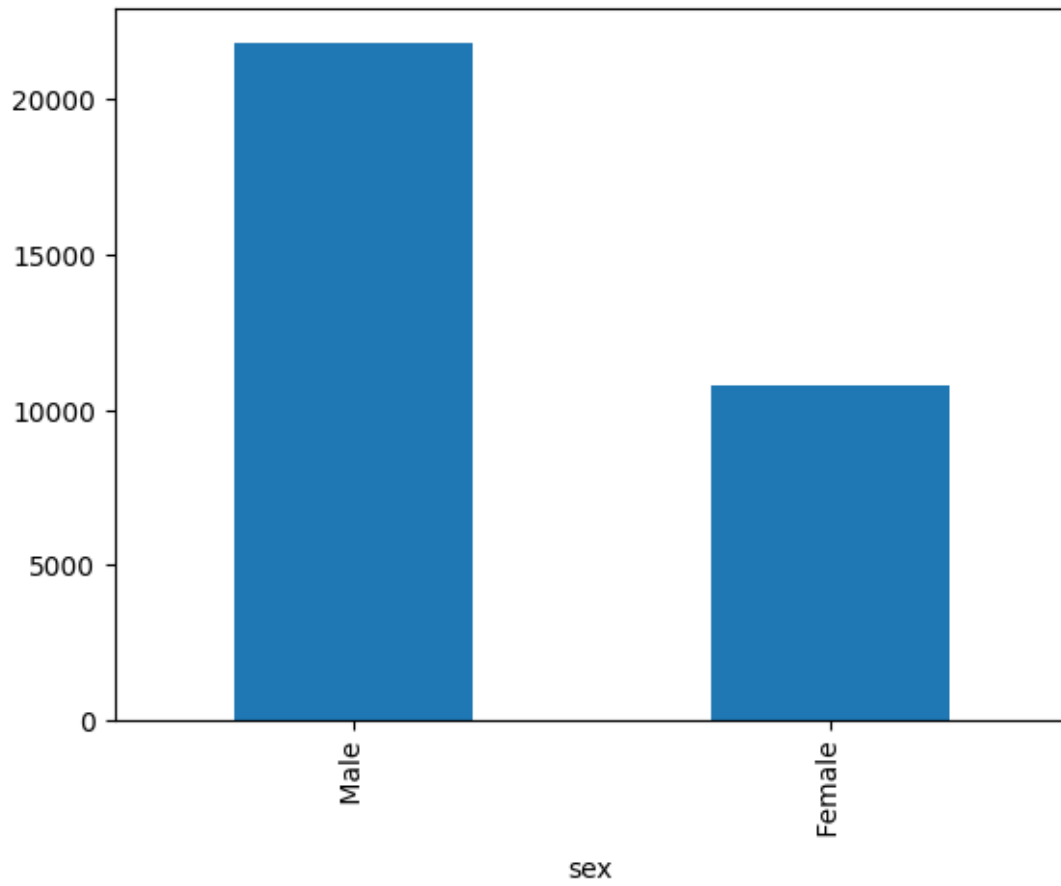
```
[27]: tab_fr_sex=data["sex"].value_counts()
```

```
[28]: tab_fr_sex
```

```
[28]: sex
      Male      21790
      Female  10771
      Name: count, dtype: int64
```

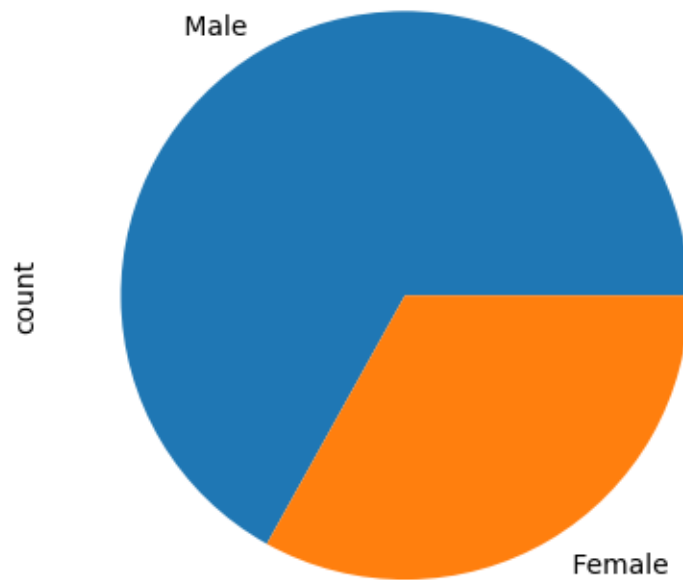
```
[29]: tab_fr_sex.plot.bar()
```

```
[29]: <Axes: xlabel='sex'>
```



```
[30]: tab_fr_sex.plot.pie()
```

```
[30]: <Axes: ylabel='count'>
```



Filtrar a los empleados con sueldo mayor a 50k para crear nuevos gráficos

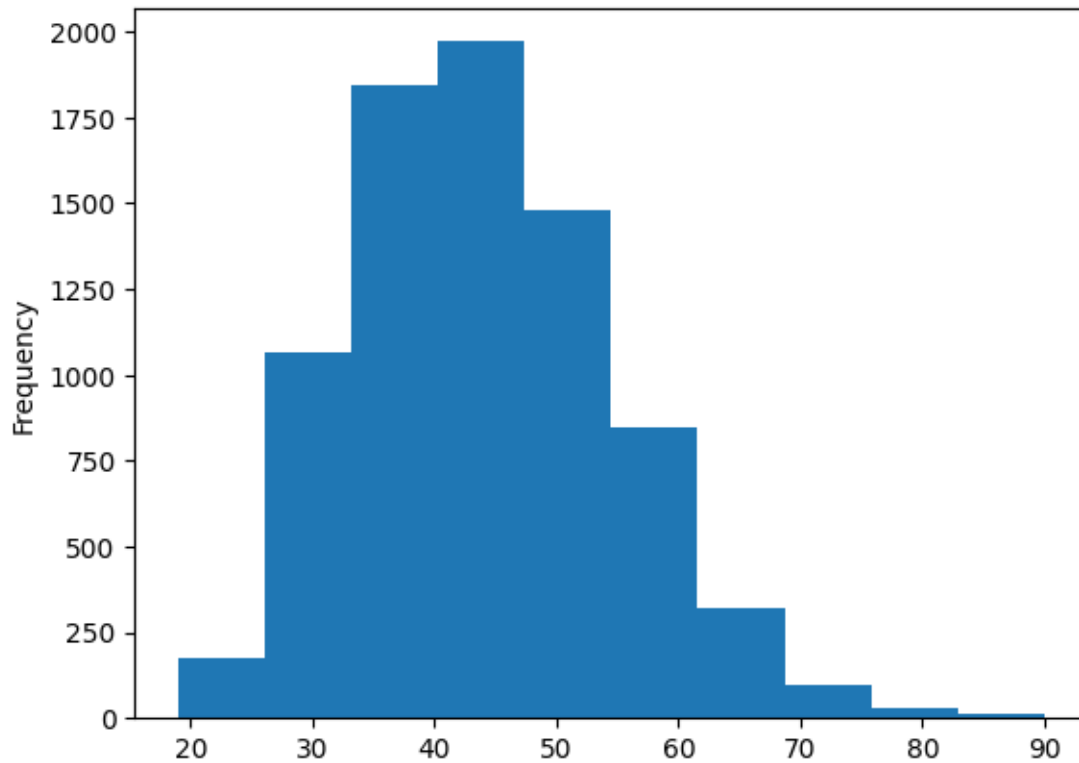
```
[31]: sal_mayor50=data[data["salary"]==" >50K"]
```

```
[32]: sal_mayor50.shape
```

```
[32]: (7841, 15)
```

```
[33]: sal_mayor50["age"].plot.hist()
```

```
[33]: <Axes: ylabel='Frequency'>
```



```
[34]: tab_fr_sex_mayor50=sal_mayor50["sex"].value_counts()  
      tab_fr_sex_mayor50
```

```
[34]: sex  
      Male      6662  
      Female   1179  
      Name: count, dtype: int64
```

```
[35]: tab_fr_sex_mayor50.plot.pie()
```

```
[35]: <Axes: ylabel='count'>
```

