

Evidence of Discrimination?

The Department of Developmental Services (DDS) in California is responsible for allocating funds to support over 250,000 developmentally-disabled residents. The data set `ca_dds_expenditures.csv` contains data about 1,000 of these residents. The data comes from a discrimination lawsuit which alleged that California's Department of Developmental Services (DDS) privileged white (non-Hispanic) residents over Hispanic residents in allocating funds. We will focus on comparing the allocation of funds (i.e., expenditures) for these two ethnicities only, although there are other ethnicities in this data set.

There are 6 variables in this data set:

- Id: 5-digit, unique identification code for each consumer (similar to a social security number and used for identification purposes)
- Age Cohort: Binned age variable represented as six age cohorts (0-5, 6-12, 13-17, 18-21, 22-50, and 51+)
- Age: Unbinned age variable
- Gender: Male or Female
- Expenditures: Dollar amount of annual expenditures spent on each consumer
- Ethnicity: Eight ethnic groups (American Indian, Asian, Black, Hispanic, Multi-race, Native Hawaiian, Other, and White non-Hispanic)

Question 1

Read in the data set. Make a graphic that compares the *average* expenditures by the DDS on Hispanic residents and white (non-Hispanic) residents. Comment on what you see.

In [1]:

```
%matplotlib inline
import numpy as np
import pandas as pd
```

In [2]:

```
dds_data = pd.read_csv("ca_dds_expenditures.csv")
dds_data
```

Out[2]:

	Id	Age Cohort	Age	Gender	Expenditures	Ethnicity
0	10210	13 to 17	17	Female	2113	White not Hispanic
1	10409	22 to 50	37	Male	41924	White not Hispanic
2	10486	0 to 5	3	Male	1454	Hispanic
3	10538	18 to 21	19	Female	6400	Hispanic
4	10568	13 to 17	13	Male	4412	White not Hispanic
...
995	99622	51+	86	Female	57055	White not Hispanic
996	99715	18 to 21	20	Male	7494	Hispanic
997	99718	13 to 17	17	Female	3673	Multi Race
998	99791	6 to 12	10	Male	3638	Hispanic
999	99898	22 to 50	23	Male	26702	White not Hispanic

1000 rows × 6 columns

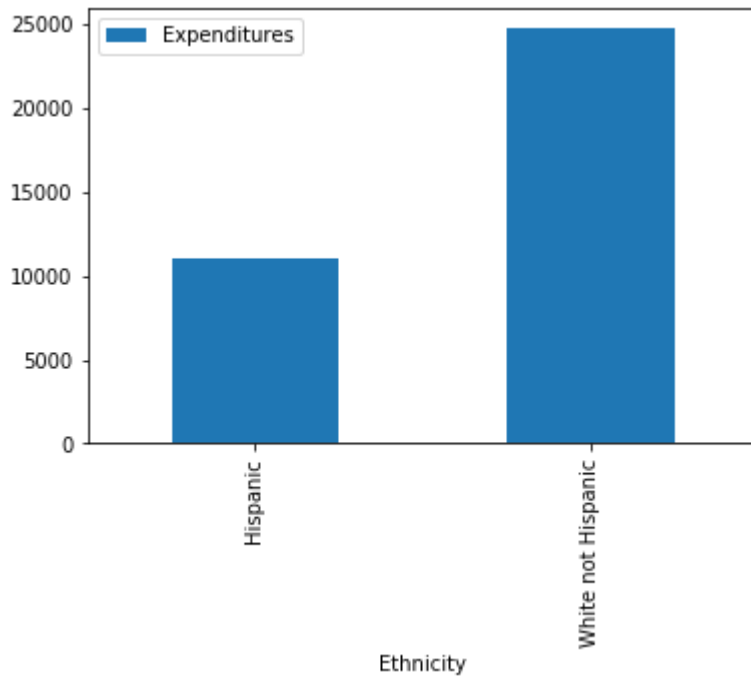
Create the cube with index Ethnicity and values average Expenditures
 Then plot the graph for only 2 category: Hispanic and Whites

In [3]:

```
expend_cube = dds_data.pivot_table(index = "Ethnicity", values = "Expenditures", aggfunc =  
np.mean)  
expend_cube.loc[["Hispanic", "White not Hispanic"]].plot.bar()
```

Out[3]:

<matplotlib.axes._subplots.AxesSubplot at 0x26cc36f7088>

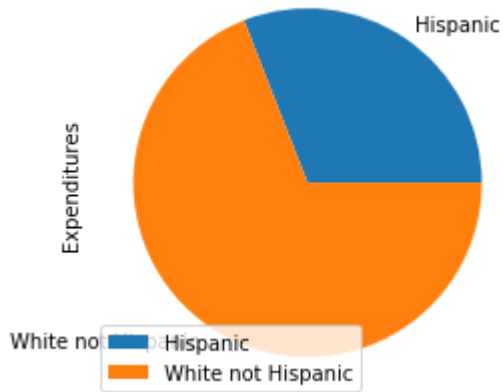


In [4]:

```
expend_cube.loc[["Hispanic", "White not Hispanic"]].plot.pie(subplots=True)
```

Out[4]:

```
array([<matplotlib.axes._subplots.AxesSubplot object at 0x0000026CC3A48E88>],  
      dtype=object)
```



As we can see from the bar and pie charts, Department of Developmental Services (DDS) spends more money for 'White not Hispanic' than for 'Hispanic'.

However, let's try to plot a different way as shown below.

Question 2

Now, calculate the average expenditures by ethnicity and age cohort. Make a graphic that compares the average expenditure on Hispanic residents and white (non-Hispanic) residents, *within each age cohort*.

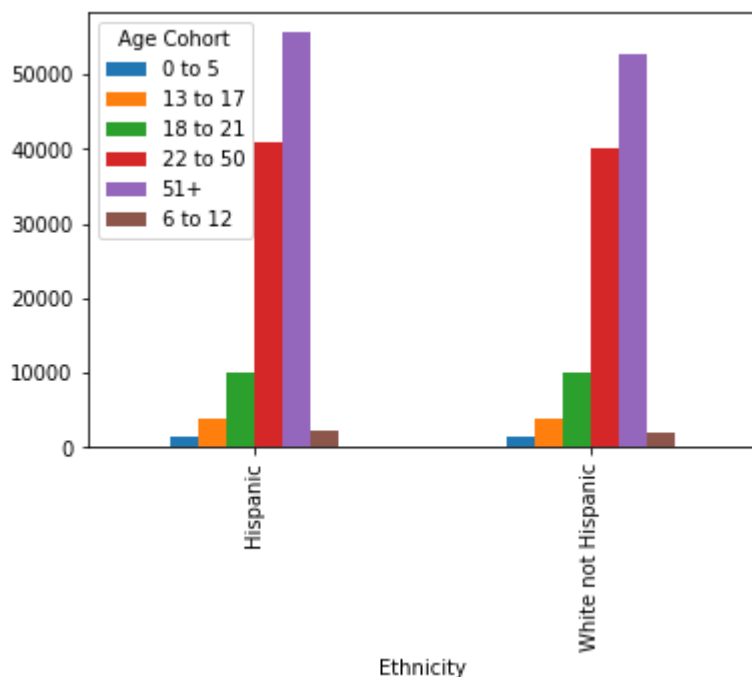
Comment on what you see. How do these results appear to contradict the results you obtained in Question 1?

In [5]:

```
expend_cube = dds_data.pivot_table(index = "Ethnicity", columns = "Age Cohort", values =
"Expenditures", aggfunc = np.mean)
expend_cube.loc[["Hispanic", "White not Hispanic"]].plot.bar()
```

Out[5]:

<matplotlib.axes._subplots.AxesSubplot at 0x26cc3a53488>



This graph contradicts our assumption of Department of Developmental Services (DDS) spending more money for 'White not Hispanic' than for 'Hispanic'.

Question 1 graphs are deceiving.

There are some biased in question 1 graphs.

This graph shows that there's no biased on white race.

Question 3

Can you explain the discrepancy between the two analyses you conducted above (i.e., Questions 1 and 2)? Try to tell a complete story that interweaves tables, graphics, and explanation.

Hint: You might want to consider looking at:

- the distributions of ages of Hispanics and whites
- the average expenditure as a function of age

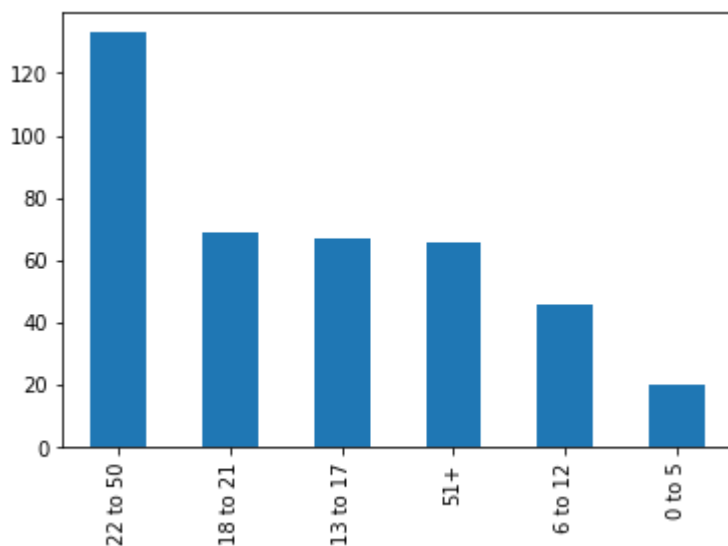
Distribution among age groups within White not Hispanic

In [6]:

```
dds_data.loc[dds_data["Ethnicity"] == "White not Hispanic", "Age Cohort"].value_counts().plot.bar()
```

Out[6]:

<matplotlib.axes._subplots.AxesSubplot at 0x26cc3b77588>



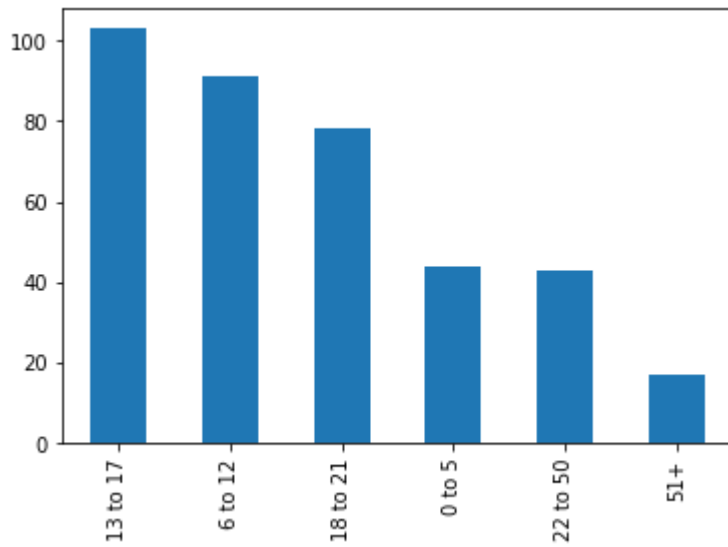
Distribution among age groups within Hispanic

In [7]:

```
dds_data.loc[dds_data["Ethnicity"] == "Hispanic", "Age Cohort"].value_counts().plot.bar()
```

Out[7]:

<matplotlib.axes._subplots.AxesSubplot at 0x26cc3bfba48>



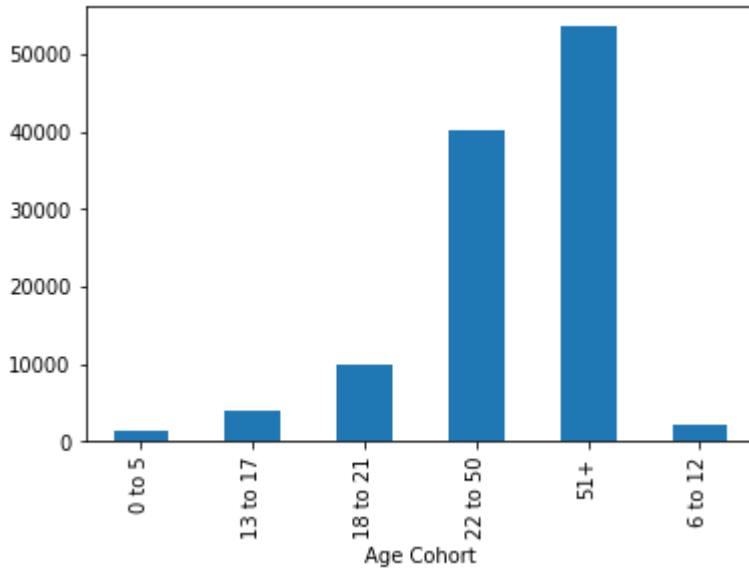
Average Expenditures for each age group despite race

In [8]:

```
dds_data.groupby("Age Cohort")["Expenditures"].mean().plot.bar()
```

Out[8]:

<matplotlib.axes._subplots.AxesSubplot at 0x26cc3c6fbc8>



Expenditures spent for 51+ have high volume.

Hispanic has only 17 people in this category and White has approximately 66.

The second highest spent group 22 to 50.

Hispanic has approximately 43 people and White has approximately 133.

They may have the same mean, but the total amounts of money for these categories are different.

Submission Instructions

Once you are finished, follow these steps:

1. Restart the kernel and re-run this notebook from beginning to end by going to `Kernel > Restart Kernel` and `Run All Cells`.
2. If this process stops halfway through, that means there was an error. Correct the error and repeat Step 1 until the notebook runs from beginning to end.
3. Double check that there is a number next to each code cell and that these numbers are in order.

Then, submit your lab as follows:

1. Go to `File > Export Notebook As > PDF`.
2. Double check that the entire notebook, from beginning to end, is in this PDF file. (If the notebook is cut off, try first exporting the notebook to HTML and printing to PDF.)
3. Upload the PDF and Notebook (ipynb) iLearn.