

CMSE201_ExampleProject_3

November 10, 2019

0.0.1 ORGANIC VS. CONVENTIONAL PRODUCTS

0.0.2 Background and Motivation

After doing much of my own personal research, it was easy for me to choose this as my semester project. I have decided to try and answer a few questions about the differences between organic and conventional food products; such as nutritional facts, ingredients, process, and prices. The specific questions I came up with were: 1. What are the nutritional differences? 2. What are the ingredient differences? 3. What are the cost differences? And does this have an effect on how much organic food is sold?

0.0.3 Methodology

```
[1]: %matplotlib inline #include all imports
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
```

The following cells were used to create bar charts of different products that are popular items for people to choose as an organic option. The bar charts are comparing the macros of organic vs. conventional. I chose to only look at the macros because after doing a lot of research on nutrition, I have found macro counting to be more effective and more important than counting calories.

```
[4]: organic_chicken = pd.read_csv("Organic chicken.csv") #open organic chicken file
      #that contains its nutritional facts
      chicken = pd.read_csv("Chicken.csv") #do the same for regular chicken
```

```
[5]: #going to find all the nutritional info for macros (protein,fats,carbs) because
      #this is what I look for on labels
```

```
organic_protein = organic_chicken.loc[2] #get row that contains organic chicken
      #protein info
the_organic_protein = organic_protein[-1] #get specific protein value

#These two steps are repeated for the organic chicken's fat and carb info

organic_fat = organic_chicken.loc[3]
the_organic_fat = organic_fat[-1]
```

```

organic_carbs = organic_chicken.loc[4]
the_organic_carbs = organic_carbs[-1]

protein = chicken.loc[2] #get regular chicken protein info & repeat for fats
→and carbs
fat = chicken.loc[3]
carbs = chicken.loc[4]

nutrients = [organic_protein[0], organic_fat[0], organic_carbs[0]]
#this creates a list containing the words "protein", "fat", "carbs" in order to
→use it for the ticks of the x axis

link = [the_organic_protein, the_organic_fat, the_organic_carbs] #put the
→organic values in a list
link2 = [protein[1], fat[1], carbs[1]] #put the regular values in a list

```

```

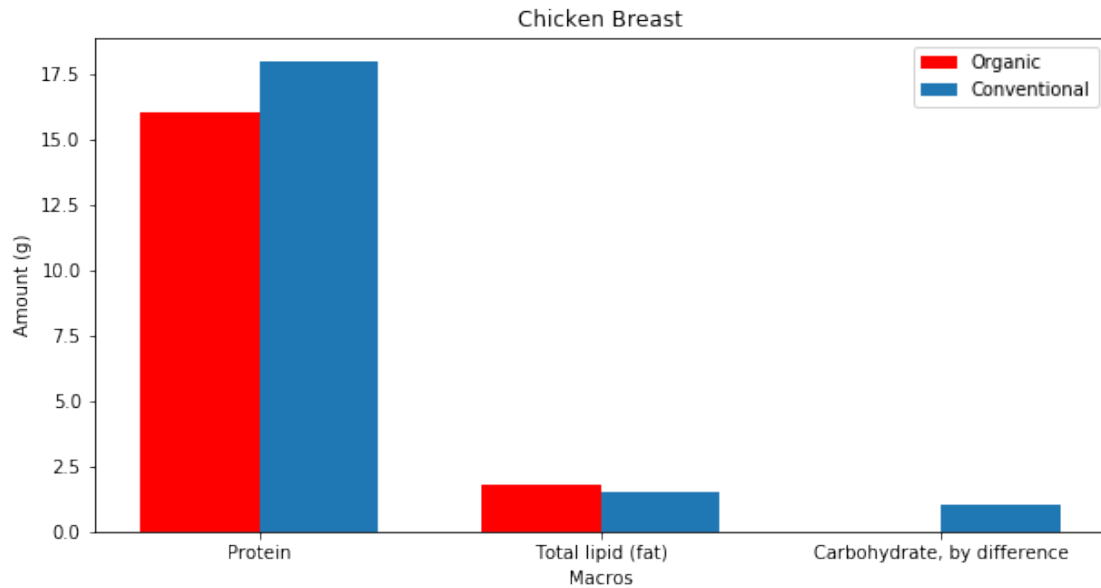
[6]: plt.figure(figsize=(10,5)) #set figure size
n_groups = 3 #there are three different groups of bars that will be on the
→chart
index = np.arange(n_groups) #the index and width came in handy when I had to
→put the bars side by side
width = 0.35

plt.bar(index, link, width, label="Organic", color='r') #graph the organic info
plt.bar(index + width, link2, width, label = "Conventional") #graph the
→conventional info

plt.xticks(index + width / 2, nutrients)
plt.xlabel("Macros")
plt.ylabel("Amount (g)")
plt.title("Chicken Breast")
plt.legend()
plt.show()

##### THIS PROCESS WAS REPEATED 2 MORE TIMES FOR THE FOLLOWING ITEMS: ICE
→CREAM SANDWICH & EGG #####

```



The next following cells are printing out the ingredient lists of organic and conventional item. I decided to look at this because this is a very important step in understanding the differences between organic and conventional. Organic products have many less ingredients, and all ingredients tend to also be organic or natural. Whereas conventional products tend to have ongoing ingredient lists, usually full of stuff I have no idea what it is and can barely pronounce. They also have many artificial and processed ingredients. I know many people do not even bother looking at the ingredient list so I found this to be an interesting and useful step.

```
[23]: my_string = organic_chicken["Nutrient"].loc[21] #Locate the row with organic
      →ingredient info
      my_list = my_string.split(",") #split the string to get items separated by
      →commas
      my_new_list = my_list[0:-1]
      last_ingredient = my_list[-1]
      new_last_ingredient = last_ingredient.split(".") #these lines were needed to
      →get rid of unnecessary info after the ingredients
      my_new_list.append(new_last_ingredient[0])
      for i, value in enumerate(my_new_list, 1): #print out the enumerated list of
      →ingredients
      print(i, value)
```

```
1 ORGANIC CHICKEN BREAST MEAT WITH RIB MEAT
2 WATER
3 SEA SALT
```

```
[24]: my_string = chicken["Nutrient"].loc[21] #Locate the row with conventional
      →ingredient info
```

```

my_list = my_string.split(",") #split the string to get items separated by
    ↳commas
my_new_list = my_list[0:-1]
last_ingredient = my_list[-1]
new_last_ingredient = last_ingredient.split(".") #these lines were needed to
    ↳get rid of unnecessary info after the ingredients
my_new_list.append(new_last_ingredient[0])
for i, value in enumerate(my_new_list, 1): #print out the enumerated list of
    ↳ingredients
    print(i, value)

    ##### THIS PROCESS WAS REPEATED AGAIN TO GET THE INGREDIENTS FOR ICE CREAM
    ↳SANDWICH #####

```

```

1 CHICKEN BREAST
2 WATER
3 SEASONING (DEHYDRATED GARLIC
4 SUGAR
5 DEHYDRATED ONION
6 SPICES
7 PAPRIKA
8 NATURAL FLAVOR [MALTODEXTRIN]
9 DEHYDRATED RED BELL PEPPER
10 DEHYDRATED SHALLOTS
11 CHICKEN BROTH
12 PARSLEY
13 OLIVE OIL)
14 CONTAINS 2% OR LESS OF SEA SALT
15 VINEGAR POWDER*
16 SEASONING (YEAST EXTRACT
17 NATURAL FLAVOR)
18 NATURAL FLAVORS

```

```

[25]: organic_ice_cream = pd.read_csv("Organic Ice Cream Sandwich.csv")
      ice_cream = pd.read_csv("Ice Cream Sandwich.csv")

```

```

[26]: organic_protein = organic_ice_cream.loc[2]
      organic_fat = organic_ice_cream.loc[3]
      organic_carbs = organic_ice_cream.loc[4]

      protein = ice_cream.loc[2]
      fat = ice_cream.loc[3]
      carbs = ice_cream.loc[4]

      nutrients = [organic_protein[0], organic_fat[0], organic_carbs[0]]

```

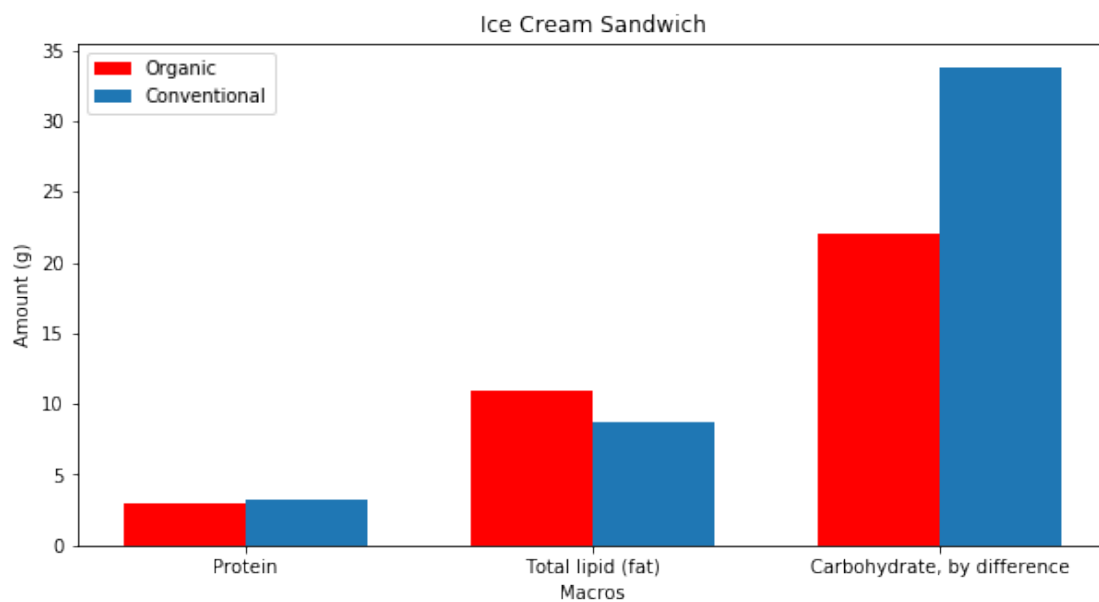
```
link = [organic_protein[1], organic_fat[1], organic_carbs[1]]

link2 = [protein[1], fat[1], carbs[1]]
```

```
[27]: plt.figure(figsize=(10,5))
n_groups = 3
index = np.arange(n_groups)
width = 0.35

plt.bar(index, link, width, label="Organic", color='r')
plt.bar(index + width, link2, width, label="Conventional")

plt.xticks(index + width / 2, nutrients)
plt.xlabel("Macros")
plt.ylabel("Amount (g)")
plt.title("Ice Cream Sandwich")
plt.legend()
plt.show()
```



```
[28]: my_string = organic_ice_cream["Nutrient"].loc[21]
my_list = my_string.split(",")
my_new_list = my_list[0:-1]
last_ingredient = my_list[-1]
new_last_ingredient = last_ingredient.split(".")
my_new_list.append(new_last_ingredient[0])
for i, value in enumerate(my_new_list, 1):
    print(i, value)
```

```

1 ICE CREAM: ORGANIC CREAM
2 ORGANIC MILK
3 ORGANIC CANE SUGAR
4 ORGANIC EGG YOLKS
5 ORGANIC VANILLA EXTRACT
6 ORGANIC LOCUST BEAN GUM
7 ORGANIC GUAR GUM.
8 CHOCOLATE WAFER: ORGANIC WHEAT FLOUR
9 ORGANIC SUGAR
10 ORGANIC CANE SYRUP
11 ORGANIC PALM OIL
12 ORGANIC CARAMEL COLOR
13 BAKING SODA
14 SOY LECITHIN
15 ORGANIC COCOA
16 SALT
17 NATURAL VANILLA FLAVOR

```

```

[29]: my_string = ice_cream["Nutrient"].loc[21]
      my_list = my_string.split(",")
      my_new_list = my_list[0:-1]
      last_ingredient = my_list[-1]
      new_last_ingredient = last_ingredient.split(".")
      my_new_list.append(new_last_ingredient[0])
      for i, value in enumerate(my_new_list, 1):
          print(i, value)

```

```

1 ICE CREAM - MILK
2 CREAM
3 CORN SYRUP
4 SUGAR
5 WHEY
6 BUTTERMILK
7 MALTODEXTRIN
8 CELLULOSE GEL
9 MONO AND DIGLYCERIDES
10 GUAR GUM
11 NATURAL AND ARTIFICIAL FLAVORS
12 LOCUST BEAN GUM
13 CELLULOSE GUM
14 POLYSORBATE 80
15 CARRAGEENAN
16 SUCRALOSE. WAFERS - BLEACHED WHEAT FLOUR
17 SUGAR
18 WHOLE WHEAT FLOUR
19 PALM OIL
20 DEXTROSE

```

```

21  MOLASSES
22  HIGH FRUCTOSE CORN STARCH
23  SALT
24  BAKING SODA
25  NATURAL FLAVOR
26  MONO & DIGLYCERIDES
27  CARAMEL COLOR
28  SOY LECITHIN. CHOCO CHIPS - SUGAR
29  COCONUT OIL
30  COCOA PROCESSES WITH ALKALI
31  PARTIALLY HYDROGENATED COCONUT OIL
32  COCOA
33  SALT
34  LECITHIN
35  NATURAL FLAVOR

```

```

[30]: organic_egg = pd.read_csv("Organic Egg.csv")
      egg = pd.read_csv("Egg.csv")

```

```

[31]: organic_protein = organic_egg.loc[2]
      organic_fat = organic_egg.loc[3]
      organic_carbs = organic_egg.loc[4]

      protein = egg.loc[3]
      fat = egg.loc[4]
      carbs = egg.loc[5]

      nutrients = [organic_protein[0], organic_fat[0], organic_carbs[0]]
      link = [organic_protein[1], organic_fat[1], organic_carbs[1]]

      link2 = [protein[2], fat[2], carbs[2]]

```

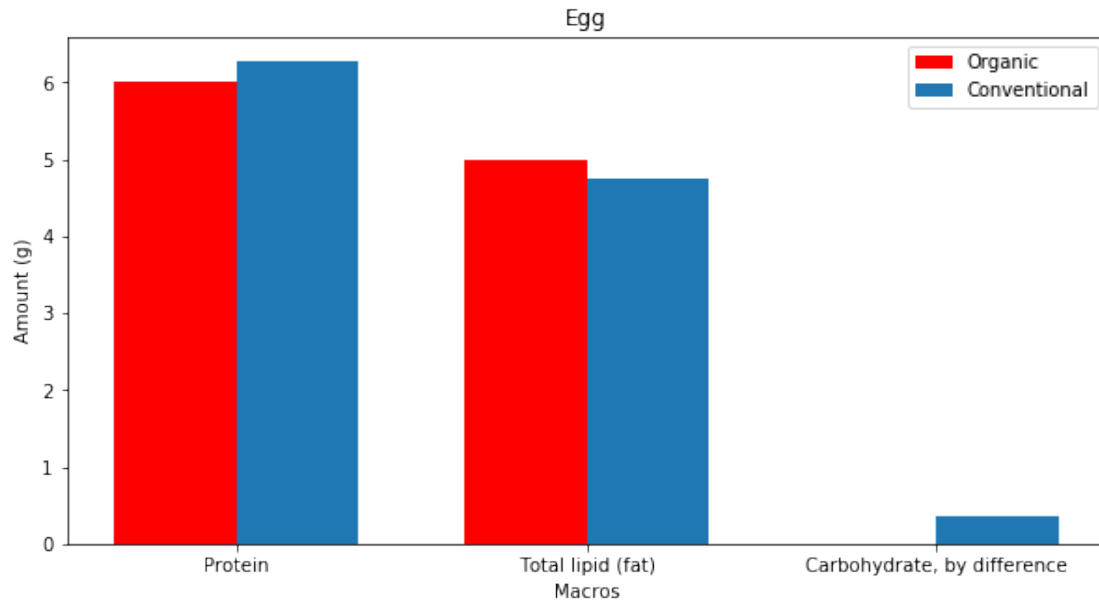
```

[32]: plt.figure(figsize=(10,5))
      n_groups = 3
      index = np.arange(n_groups)
      width = 0.35

      plt.bar(index, link, width, label="Organic",color='r')
      plt.bar(index + width, link2, width, label="Conventional")

      plt.xticks(index + width / 2, nutrients)
      plt.xlabel("Macros")
      plt.ylabel("Amount (g)")
      plt.title("Egg")
      plt.legend()
      plt.show()

```



Fruits and vegetables have always been at the top of organic food sales, and some of the most popular products people should always buy organic are apples, strawberries, and lettuce. These next cells show how the price of these items have changed over time, and I predicted the price of them for the next year.

```
[57]: apples = pd.read_csv("Apple prices.csv") #open file
apple_date = apples["REPORT_DATE"] #get specific columns
apple_price = apples["WEIGHTED_AVERAGE_PRICE"]

[58]: date = np.arange(0, len(apple_date))
max_date = max(date)
extended_date = np.arange(max_date, max_date+52, 1) #create extended date array
    ↳ used 52 because there are 52 weeks in a year

parameters1 = np.polyfit(date, apple_price, 4) #use polyfit to create a
    ↳ regression line for the price data I had
my_poly_function1 = np.poly1d(parameters1)
expected_y_poly1 = my_poly_function1(date)

parameters2 = np.polyfit(date, apple_price, 4) #use polyfit to create a
    ↳ regression line predicting the values for the next year
my_poly_function2 = np.poly1d(parameters2)
expected_y_poly2 = my_poly_function1(extended_date)

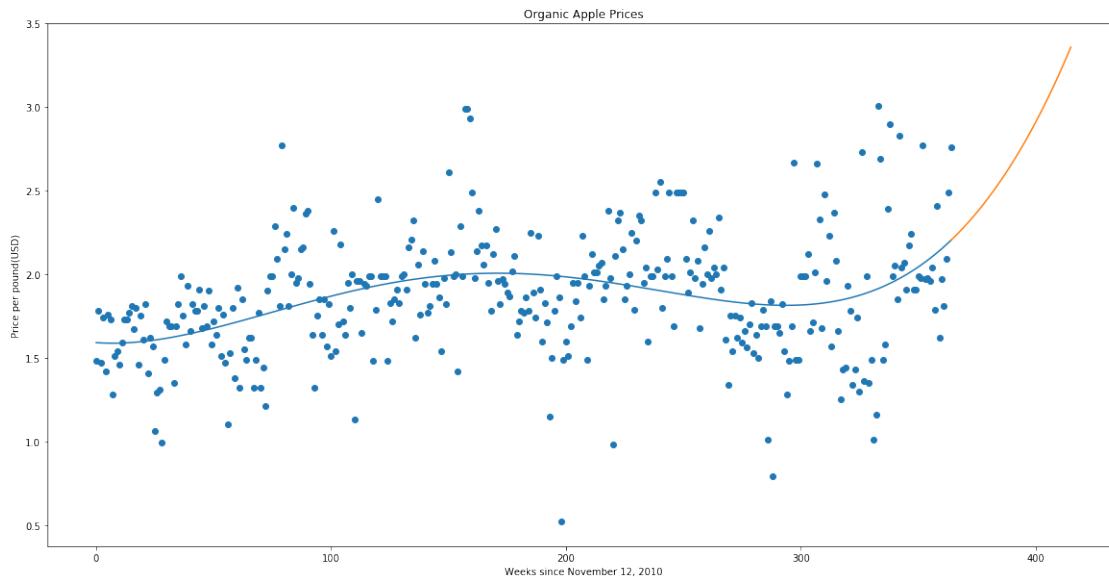
[59]: plt.figure(figsize=(20,10))
plt.xlabel("Weeks since November 12, 2010") #set labels
plt.ylabel("Price per pound(USD)")
plt.title("Organic Apple Prices")
plt.scatter(date, apple_price) #create plots and regression lines
```



```
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
```

THIS PROCESS WAS REPEATED FOR STRAWBERRY AND ROMAINE LETTUCE PRICES

[59]: [matplotlib.lines.Line2D at 0x11c3a6ba8>]



```
[61]: strawberry = pd.read_csv("Strawberry prices.csv")
strawberry_date = strawberry["REPORT_DATE"]
strawberry_price = strawberry["WEIGHTED_AVERAGE_PRICE"]
```

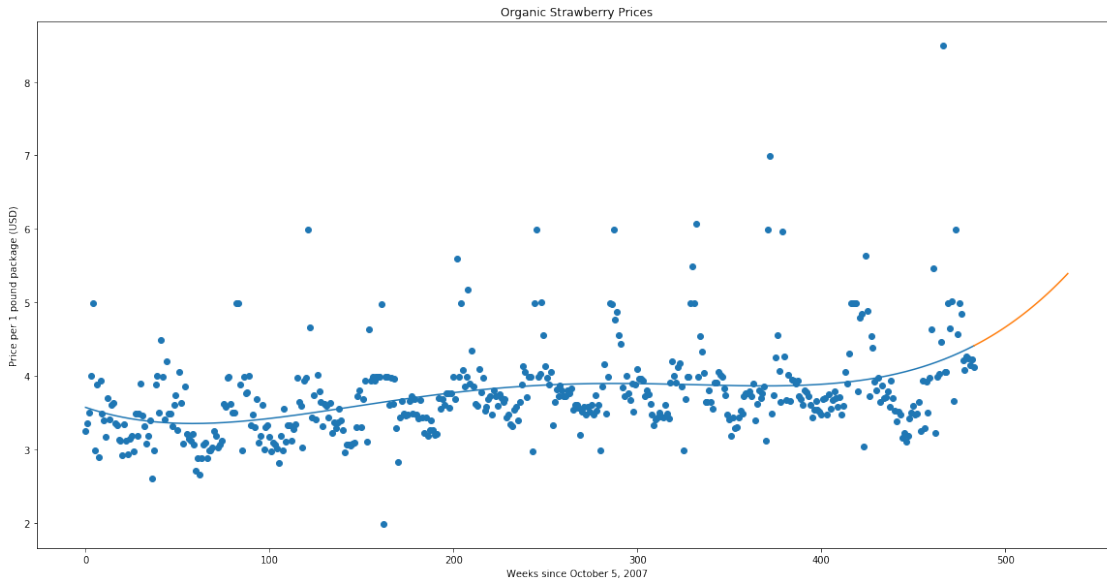
```
[62]: date = np.arange(0, len(strawberry_date))
max_date = max(date)
extended_date = np.arange(max_date, max_date+52, 1)

parameters1 = np.polyfit(date, strawberry_price, 4)
my_poly_function1 = np.poly1d(parameters1)
expected_y_poly1 = my_poly_function1(date)

parameters2 = np.polyfit(date, strawberry_price, 4)
my_poly_function2 = np.poly1d(parameters2)
expected_y_poly2 = my_poly_function1(extended_date)
```

```
[63]: plt.figure(figsize=(20,10))
plt.scatter(date, strawberry_price)
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Strawberry Prices")
plt.xlabel("Weeks since October 5, 2007")
plt.ylabel("Price per 1 pound package (USD)")
```

[63]: Text(0,0.5,'Price per 1 pound package (USD)')



```
[65]: lettuce = pd.read_csv("Lettuce prices.csv")
lettuce_date = lettuce["REPORT_DATE"]
lettuce_price = lettuce["WEIGHTED_AVERAGE_PRICE"]
```

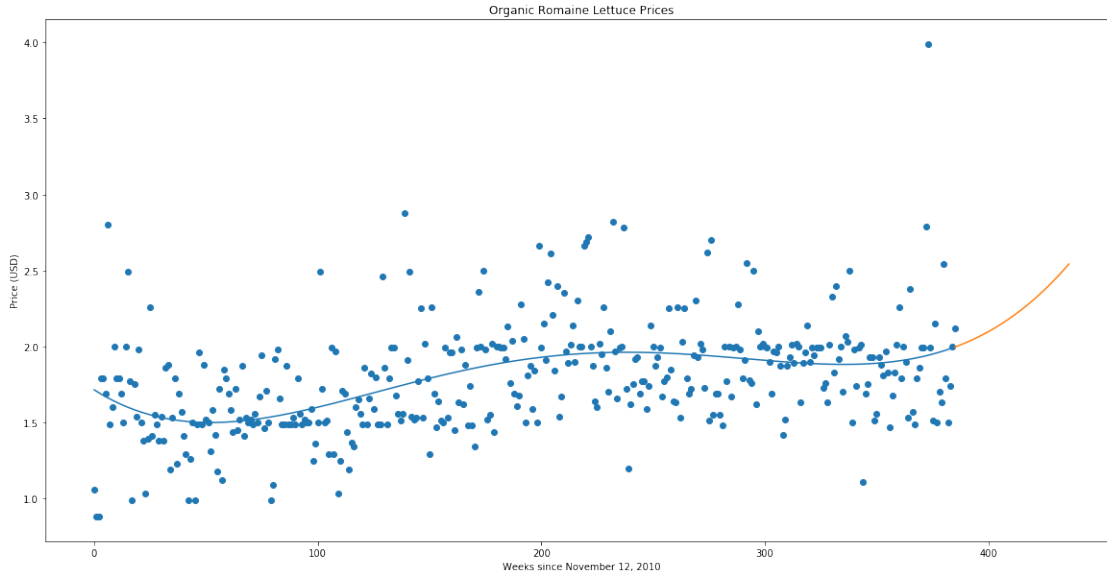
```
[66]: date = np.arange(0, len(lettuce_date))
max_date = max(date)
extended_date = np.arange(max_date, max_date+52, 1)

parameters1 = np.polyfit(date, lettuce_price, 4)
my_poly_function1 = np.poly1d(parameters1)
expected_y_poly1 = my_poly_function1(date)

parameters2 = np.polyfit(date, lettuce_price, 4)
my_poly_function2 = np.poly1d(parameters2)
expected_y_poly2 = my_poly_function1(extended_date)
```

```
[67]: plt.figure(figsize=(20,10))
plt.scatter(date, lettuce_price)
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Romaine Lettuce Prices")
plt.xlabel("Weeks since November 12, 2010")
plt.ylabel("Price (USD)")
```

[67]: Text(0,0.5,'Price (USD)')



The following cells contains data comparing prices between conventional and organic products at four different supermarkets. I was able to use masking to find the percentage differences of the same food products I used above: apples, strawberries, lettuce. I also found the overall average of how much extra we pay for organic at these supermarkets.

```
[44]: price_diff = pd.read_csv("Price Differences.csv")
price_diff
```

```
[44]:
```

	Table 2	Price Chopper	Safeway	Walmart	\
0	Apples (lb.)	NaN	NaN	NaN	
1	Regular	\$1.00	\$1.83	NaN	
2	Organic	\$1.20	\$2.20	NaN	
3	% difference	20%	20%	NaN	
4	NaN	NaN	NaN	NaN	
5	Bananas (lb.)	NaN	NaN	NaN	
6	Regular	59 cents	48 cents	58 cents	
7	Organic	79 cents	79 cents	78 cents	
8	% difference	34%	65%	34%	
9	NaN	NaN	NaN	NaN	
10	Beef (85% lean ground, lb.)	NaN	NaN	NaN	
11	Regular	NaN	\$5.99	NaN	
12	Organic	NaN	\$8.79	NaN	
13	% difference	NaN	47%	NaN	
14	NaN	NaN	NaN	NaN	
15	Butter (lb.)	NaN	NaN	NaN	
16	Regular	\$2.99/lb.	NaN	\$3.88/lb.	
17	Organic	\$7.98/lb.	NaN	\$6.48/lb.	
18	% difference	167%	NaN	67%	
19	NaN	NaN	NaN	NaN	
20	Carrots (baby, lb.)	NaN	NaN	NaN	

21	Regular	\$1.33	\$2.19	\$1.68
22	Organic	\$1.99	\$2.19	\$3.48
23	% difference	50%	0%	107%
24	NaN	NaN	NaN	NaN
25	Chicken, whole/cutup (lb.)	NaN	NaN	NaN
26	Regular	\$1.49	\$1.99	NaN
27	Organic	\$3.49	\$2.49	NaN
28	% difference	134%	25%	NaN
29	NaN	NaN	NaN	NaN
..
45	Iceberg lettuce (head)	NaN	NaN	NaN
46	Regular	\$1.99	\$2.79	\$1.68
47	Organic	\$3.49	\$3.29	\$2.48
48	% difference	75%	18%	48%
49	NaN	NaN	NaN	NaN
50	Maple syrup (Grade A, pint)	NaN	NaN	NaN
51	Regular	\$11.99	\$11.84	\$10.21
52	Organic	\$10.65	\$16.97	\$11.84
53	% difference	-11%	43%	6%
54	NaN	NaN	NaN	NaN
55	Milk (half gallon)	NaN	NaN	NaN
56	Regular	\$2.99	\$2.69	\$2.20
57	Organic	\$3.99	\$3.49	\$3.88
58	% difference	33%	30%	76%
59	NaN	NaN	NaN	NaN
60	Olive oil (extra virgin, quart)	NaN	NaN	NaN
61	Regular	\$16.08	\$8.64	\$8.98
62	Organic	\$17.02	\$13.44	\$10.87
63	% difference	6%	56%	21%
64	NaN	NaN	NaN	NaN
65	Strawberries (lb.)	NaN	NaN	NaN
66	Regular	\$2.99	\$4.39	NaN
67	Organic	\$4.99	\$7.69	NaN
68	% difference	67%	75%	NaN
69	NaN	NaN	NaN	NaN
70	Zucchini (lb.)	NaN	NaN	NaN
71	Regular	\$1.99	72 cents	\$1.80
72	Organic	\$2.99	\$1.12	\$1.98
73	% difference	50%	56%	10%
74	Average premium for organic	59%	34%	51%

Whole Foods

0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

5	NaN
6	79 cents
7	99 cents
8	25%
9	NaN
10	NaN
11	\$6.99
12	\$9.99
13	43%
14	NaN
15	NaN
16	\$3.79/lb.
17	\$4.39.1b.
18	16%
19	NaN
20	NaN
21	NaN
22	NaN
23	NaN
24	NaN
25	NaN
26	\$2.49
27	\$3.49
28	40%
29	NaN
..	...
45	NaN
46	NaN
47	NaN
48	NaN
49	NaN
50	NaN
51	\$11.99
52	\$11.72
53	-2%
54	NaN
55	NaN
56	\$2.39
57	\$3.99
58	67%
59	NaN
60	NaN
61	\$13.24
62	\$13.24
63	0%
64	NaN
65	NaN

```

66      $4.99
67      $6.99
68      40%
69      NaN
70      NaN
71      NaN
72      NaN
73      NaN
74      24%

```

```
[75 rows x 5 columns]
```

```
[45]: price_diff.loc[0:3]
```

```

[45]:      Table 2 Price Chopper Safeway Walmart Whole Foods
0 Apples (lb.)      NaN      NaN      NaN      NaN
1      Regular      $1.00    $1.83      NaN      NaN
2      Organic      $1.20    $2.20      NaN      NaN
3 % difference      20%      20%      NaN      NaN

```

```
[46]: price_diff.loc[65:68]
```

```

[46]:      Table 2 Price Chopper Safeway Walmart Whole Foods
65 Strawberries (lb.)      NaN      NaN      NaN      NaN
66      Regular      $2.99    $4.39      NaN    $4.99
67      Organic      $4.99    $7.69      NaN    $6.99
68 % difference      67%      75%      NaN      40%

```

```
[47]: price_diff.loc[45:48]
```

```

[47]:      Table 2 Price Chopper Safeway Walmart Whole Foods
45 Iceberg lettuce (head)      NaN      NaN      NaN      NaN
46      Regular      $1.99    $2.79    $1.68      NaN
47      Organic      $3.49    $3.29    $2.48      NaN
48 % difference      75%      18%      48%      NaN

```

```
[48]: price_diff.loc[74]
```

```

[48]: Table 2      Average premium for organic
Price Chopper      59%
Safeway      34%
Walmart      51%
Whole Foods      24%
Name: 74, dtype: object

```

The following cells show how sales of organic food in the U.S. have changed over time. I was very interested in showing this after showing the price differences, because I wanted to show that sales are still increasing in spite of the costs of organic products also increasing.

```

[72]: us_sales = pd.read_csv("US Organic Sales.csv", skiprows=2)
      years = us_sales["Year"]
      money = us_sales["Sales in billion U.S. dollars"]

```

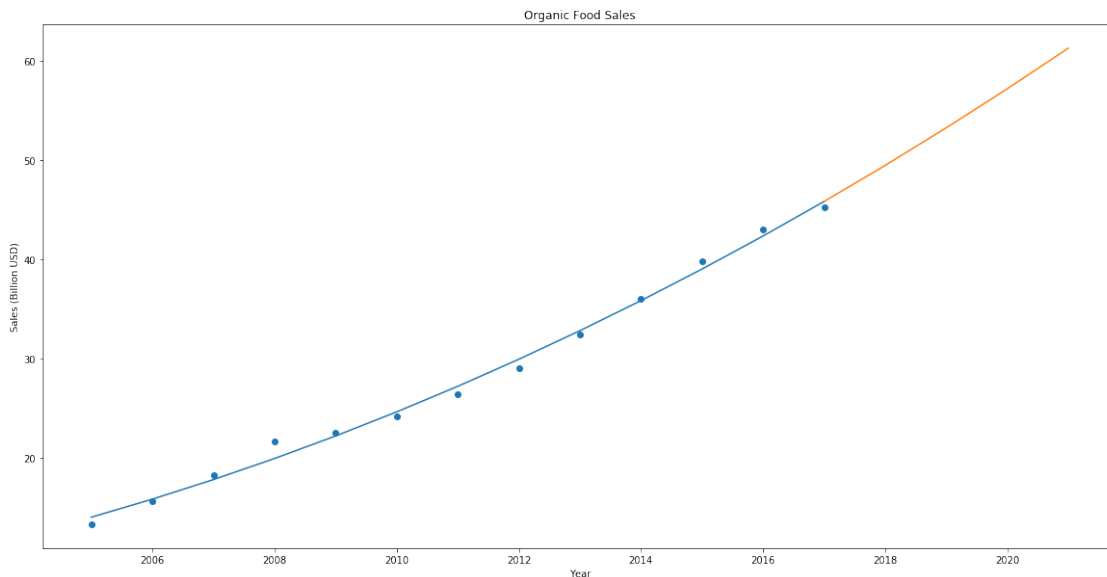
```
[73]: max_date = max(years)
extended_date = np.arange(max_date, max_date+5, 1)

parameters1 = np.polyfit(years, money, 2)
my_poly_function1 = np.poly1d(parameters1)
expected_y_poly1 = my_poly_function1(years)

parameters2 = np.polyfit(years, money, 2)
my_poly_function2 = np.poly1d(parameters2)
expected_y_poly2 = my_poly_function1(extended_date)
```

```
[52]: plt.figure(figsize=(20,10))
plt.scatter(years, money)
plt.plot(years, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Food Sales")
plt.xlabel("Year")
plt.ylabel("Sales (Billion USD)")
```

```
[52]: Text(0,0.5,'Sales (Billion USD)')
```

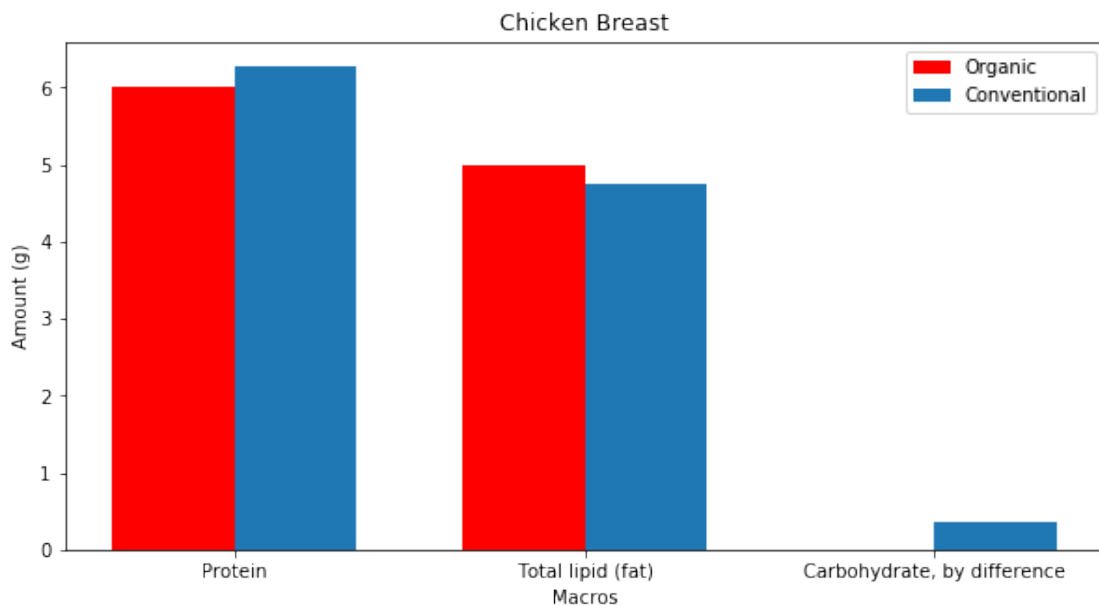


0.0.4 Results

```
[53]: plt.figure(figsize=(10,5)) #set figure size
n_groups = 3 #there are three different groups of bars that will be on the
↳chart
index = np.arange(n_groups) #the index and width came in handy when I had to
↳put the bars side by side
width = 0.35
```

```
plt.bar(index, link, width, label="Organic", color='r') #graph the organic info
plt.bar(index + width, link2, width, label = "Conventional") #graph the
    ↳conventional info

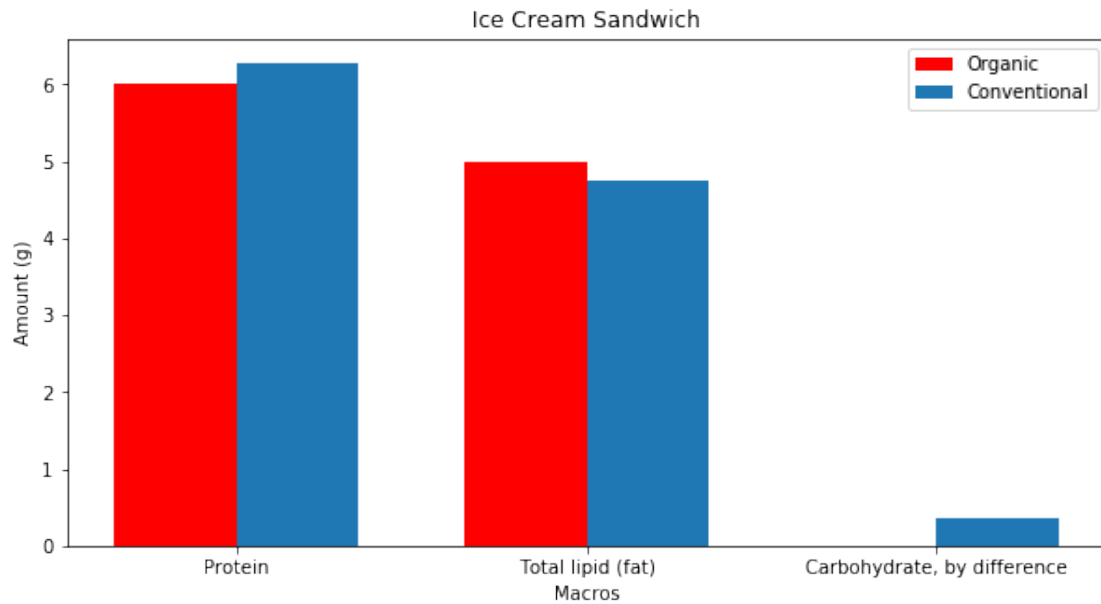
plt.xticks(index + width / 2, nutrients)
plt.xlabel("Macros")
plt.ylabel("Amount (g)")
plt.title("Chicken Breast")
plt.legend()
plt.show()
```



```
[54]: plt.figure(figsize=(10,5))
n_groups = 3
index = np.arange(n_groups)
width = 0.35

plt.bar(index, link, width, label="Organic", color='r')
plt.bar(index + width, link2, width, label="Conventional")

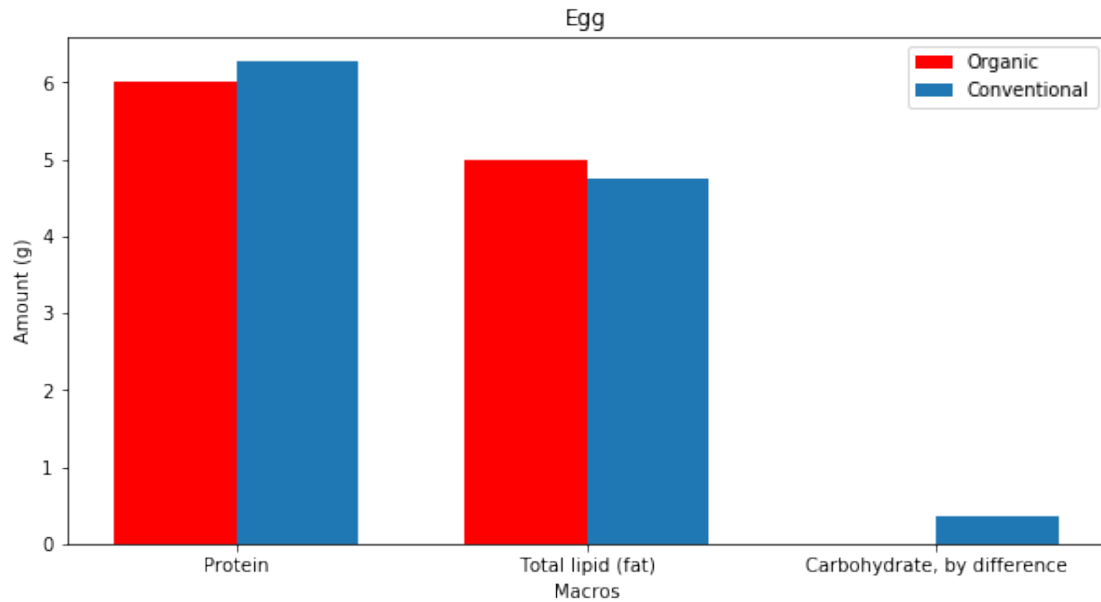
plt.xticks(index + width / 2, nutrients)
plt.xlabel("Macros")
plt.ylabel("Amount (g)")
plt.title("Ice Cream Sandwich")
plt.legend()
plt.show()
```

```
[55]: plt.figure(figsize=(10,5))
n_groups = 3
index = np.arange(n_groups)
width = 0.35

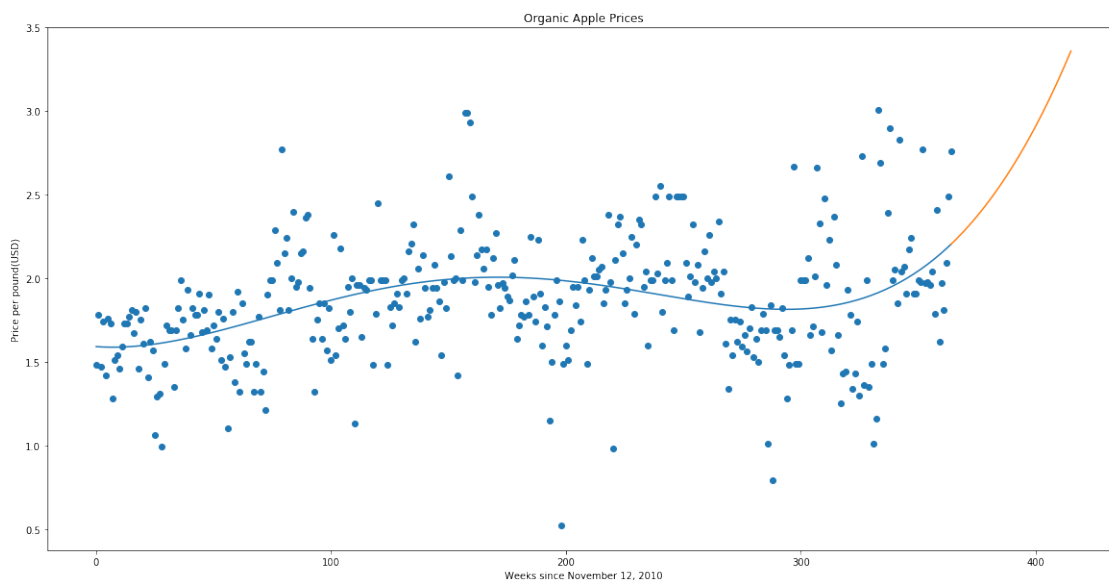
plt.bar(index, link, width, label="Organic",color='r')
plt.bar(index + width, link2, width, label="Conventional")

plt.xticks(index + width / 2, nutrients)
plt.xlabel("Macros")
plt.ylabel("Amount (g)")
plt.title("Egg")
plt.legend()
plt.show()
```



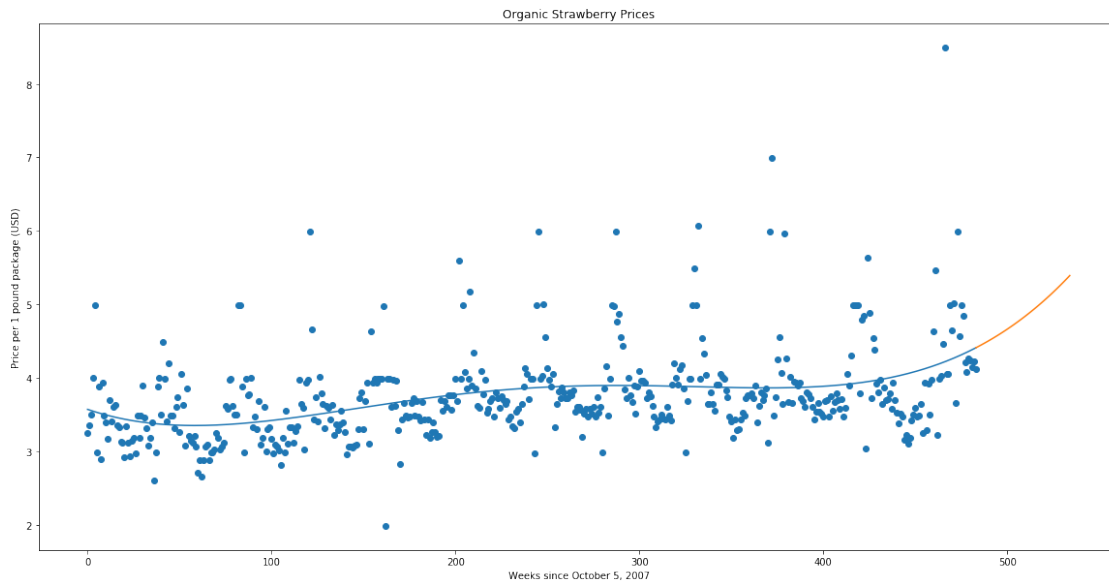
```
[60]: plt.figure(figsize=(20,10))
plt.xlabel("Weeks since November 12, 2010") #set labels
plt.ylabel("Price per pound(USD)")
plt.title("Organic Apple Prices")
plt.scatter(date, apple_price) #create plots and regression lines
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
```

[60]: [<matplotlib.lines.Line2D at 0x118f83f98>]



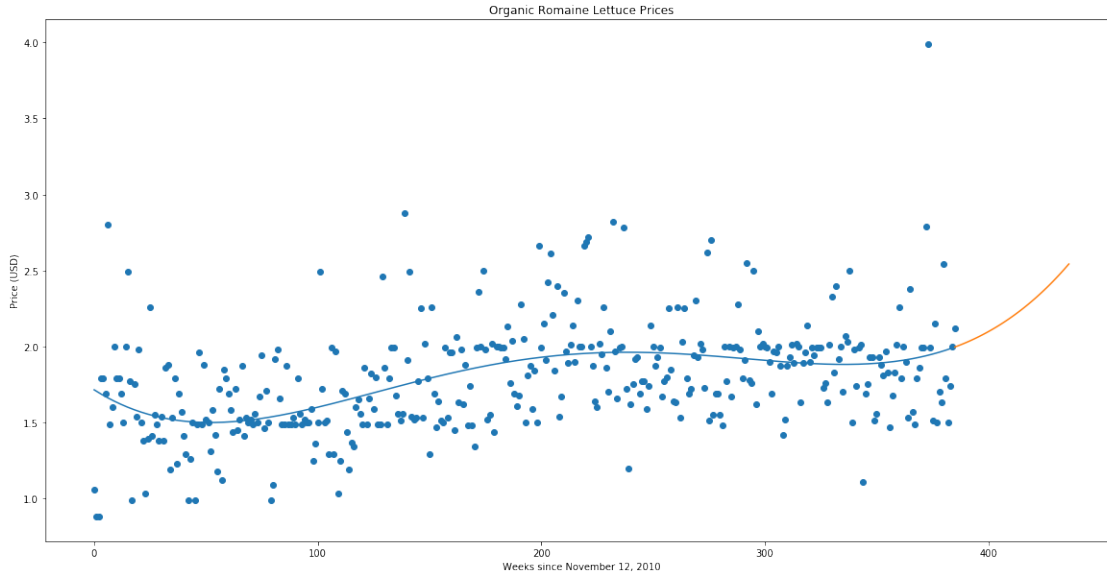
```
[64]: plt.figure(figsize=(20,10))
plt.scatter(date, strawberry_price)
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Strawberry Prices")
plt.xlabel("Weeks since October 5, 2007")
plt.ylabel("Price per 1 pound package (USD)")
```

```
[64]: Text(0,0.5,'Price per 1 pound package (USD)')
```



```
[68]: plt.figure(figsize=(20,10))
plt.scatter(date, lettuce_price)
plt.plot(date, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Romaine Lettuce Prices")
plt.xlabel("Weeks since November 12, 2010")
plt.ylabel("Price (USD)")
```

```
[68]: Text(0,0.5,'Price (USD)')
```



```
[69]: price_diff.loc[0:3]
```

```
[69]:      Table 2 Price Chopper Safeway Walmart Whole Foods
0 Apples (lb.)      NaN      NaN      NaN      NaN
1      Regular      $1.00    $1.83      NaN      NaN
2      Organic      $1.20    $2.20      NaN      NaN
3 % difference      20%     20%      NaN      NaN
```

```
[70]: price_diff.loc[65:68]
```

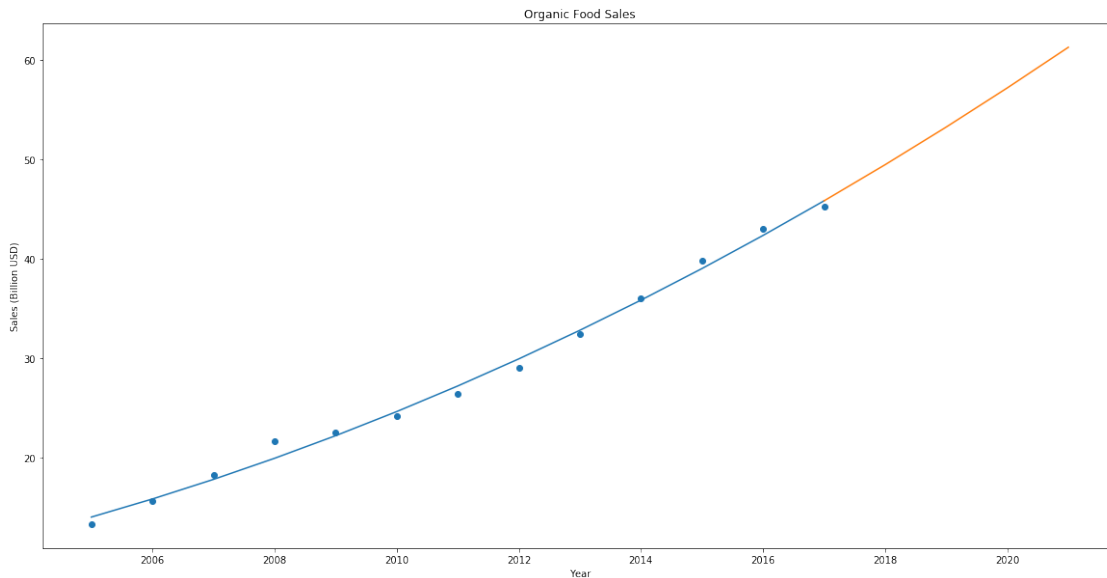
```
[70]:      Table 2 Price Chopper Safeway Walmart Whole Foods
65 Strawberries (lb.)      NaN      NaN      NaN      NaN
66      Regular      $2.99    $4.39      NaN    $4.99
67      Organic      $4.99    $7.69      NaN    $6.99
68      % difference      67%     75%      NaN     40%
```

```
[71]: price_diff.loc[45:48]
```

```
[71]:      Table 2 Price Chopper Safeway Walmart Whole Foods
45 Iceberg lettuce (head)      NaN      NaN      NaN      NaN
46      Regular      $1.99    $2.79    $1.68      NaN
47      Organic      $3.49    $3.29    $2.48      NaN
48      % difference      75%     18%     48%      NaN
```

```
[74]: plt.figure(figsize=(20,10))
plt.scatter(years, money)
plt.plot(years, expected_y_poly1)
plt.plot(extended_date, expected_y_poly2)
plt.title("Organic Food Sales")
plt.xlabel("Year")
plt.ylabel("Sales (Billion USD)")
```

[74]: Text(0,0.5,'Sales (Billion USD)')



0.05 Discussion and Conclusion

Some obstacles that I ran into with this project were: 1. I originally wanted to model the differences between how organic and conventional foods are grown, and the difference between the pesticides used. However, this became such a complex and long task because there is so much information about these and every farm grows their things differently. So, unfortunately I was unable to include this in my project. 2. It was really hard finding data on this at first, because organic products have really only become popular within the last 5-10 years.

Some things I would have done differently are maybe start everything sooner, so I am not working up until the very last minute on this. I also wish I was able to find data sets on the demographics of what types of people purchase organic.

From my results, it can be found that there are essentially no nutritional difference in a products macros (carbs, fats, proteins) between organic and conventional. There is usually a 1-2g difference between the two. Again, I only chose to look at the macros because that is what I personally look for on a label to help me determine if it is "healthy" enough to include in my diet.

Even though there are little to no nutritional differences, there are very big differences when it comes to ingredients. Conventional products have twice or more times of ingredients than organic, and the ingredients in conventional include a lot of processed or artificial things, most of which I do not know what it is nor can even pronounce. It is clear to see that people do not buy organic because of its additional nutrients, since there are none, they buy it for the things it doesn't have; such as artificial/processed ingredients and pesticides.

The last thing I decided to compare were the prices of these two food types. Fruits and vegetables are the #1 organic product bought, and so I decided to look at a few of the prices for the top fruits/vegetables that people buy organic, which are apples, strawberries, and lettuce. From the graphs it seemed that they all steadily increased in prices up until 2011-2012 and then they were all able to dip a little in prices. However, within the last 1-2 years the prices have begun to

increase, and will continually increase this next year. Within this next year apples will increase by more than \$1, strawberries will increase by \$1, and lettuce will increase by around 50 cents.

While these may not seem like huge price increases, when we compare these to the prices of the same conventional products it is shown that we are paying an extra 20% for organic apples, an extra 40-75% for organic strawberries, and an extra 18-75% for lettuce (these depend on which supermarket you shop at). On average people will pay an extra 40-50% for organic products.

I wanted to see if organic sales were going to decrease due to the expected increase in prices, however the complete opposite happened. Organic sales have been steadily increasing since 2005, and are expected to continue to increase by \$3 billion per year, for the next 5 years.

Seeing that sales in organic products are increasing drastically in spite of their prices rising, it is viable to assume that people are realizing the health benefits of organic products and believe they are worth the cost.

0.0.6 References

Food Composition Databases Show Foods List, ndb.nal.usda.gov/ndb/search/list.

"Fruits and Vegetables Top Organic Food Sales." USDA ERS - Chart Detail, www.ers.usda.gov/data-products/chartgallery/gallery/chart-detail/?chartId=87354.

"Market News Data Download." Market News, marketnews.usda.gov/mnp/dataDownload.

"Organic Food Sales in the U.S. 2017 | Statistic." Statista, www.statista.com/statistics/196952/organic-food-sales-in-the-us-since-2000/.