

PH_Food_Prices

April 1, 2024

1 PH Food Prices (2000-2023)

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

1.0.1 Read the file and analyze its data

```
[2]: # Dataset from https://data.humdata.org/dataset/wfp-food-prices-for-philippines
food_prices = pd.read_csv('ph_food_prices.csv')
food_prices.head()
```

```
[2]:
```

	date	region	city/province	market	\
0	15/01/2000	National Capital region	Metropolitan Manila	Metro Manila	
1	15/01/2000	National Capital region	Metropolitan Manila	Metro Manila	
2	15/01/2000	National Capital region	Metropolitan Manila	Metro Manila	
3	15/01/2000	National Capital region	Metropolitan Manila	Metro Manila	
4	15/01/2000	National Capital region	Metropolitan Manila	Metro Manila	

	latitude	longitude	category	commodity	unit	\
0	14.604167	120.982222	cereals and tubers	Maize flour (yellow)	KG	
1	14.604167	120.982222	cereals and tubers	Rice (milled, superior)	KG	
2	14.604167	120.982222	cereals and tubers	Rice (milled, superior)	KG	
3	14.604167	120.982222	cereals and tubers	Rice (regular, milled)	KG	
4	14.604167	120.982222	cereals and tubers	Rice (regular, milled)	KG	

	priceflag	pricetype	currency	price	usdprice
0	actual	Retail	PHP	15.00	0.3717
1	actual	Retail	PHP	20.00	0.4957
2	actual	Wholesale	PHP	18.35	0.4548
3	actual	Retail	PHP	18.00	0.4461
4	actual	Wholesale	PHP	16.35	0.4052

```
[3]: food_prices.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 141765 entries, 0 to 141764
Data columns (total 14 columns):
```

#	Column	Non-Null Count	Dtype
0	date	141765 non-null	object
1	region	141765 non-null	object
2	city/province	141765 non-null	object
3	market	141765 non-null	object
4	latitude	141765 non-null	float64
5	longitude	141765 non-null	float64
6	category	141765 non-null	object
7	commodity	141765 non-null	object
8	unit	141765 non-null	object
9	priceflag	141765 non-null	object
10	pricetype	141765 non-null	object
11	currency	141765 non-null	object
12	price	141765 non-null	float64
13	usdprice	141765 non-null	float64

dtypes: float64(4), object(10)

memory usage: 15.1+ MB

```
[4]: food_prices.describe(include='all')
```

```
[4]:
```

	date	region	city/province	market	latitude	\
count	141765	141765	141765	141765	141765.000000	
unique	281	17	78	97	NaN	
top	15/07/2020	Region III	Davao del Sur	Davao City	NaN	
freq	3906	12110	5967	4316	NaN	
mean	NaN	NaN	NaN	NaN	11.671230	
std	NaN	NaN	NaN	NaN	3.616373	
min	NaN	NaN	NaN	NaN	5.029099	
25%	NaN	NaN	NaN	NaN	8.477437	
50%	NaN	NaN	NaN	NaN	11.560250	
75%	NaN	NaN	NaN	NaN	14.608900	
max	NaN	NaN	NaN	NaN	18.194082	

	longitude	category	commodity	unit	\
count	141765.000000	141765	141765	141765	
unique	NaN	4	67	2	
top	NaN	vegetables and fruits	Rice (regular, milled)	KG	
freq	NaN	60870	6065	136845	
mean	122.881505	NaN	NaN	NaN	
std	1.924088	NaN	NaN	NaN	
min	118.735278	NaN	NaN	NaN	
25%	121.086142	NaN	NaN	NaN	
50%	122.755280	NaN	NaN	NaN	
75%	124.716536	NaN	NaN	NaN	
max	126.207645	NaN	NaN	NaN	

	priceflag	pricetype	currency	price	usdprice
count	141765	141765	141765	141765.000000	141765.000000
unique	1	3	1	NaN	NaN
top	actual	Retail	PHP	NaN	NaN
freq	141765	135111	141765	NaN	NaN
mean	NaN	NaN	NaN	116.467966	2.287652
std	NaN	NaN	NaN	109.621618	2.117082
min	NaN	NaN	NaN	1.600000	0.034300
25%	NaN	NaN	NaN	40.710000	0.810200
50%	NaN	NaN	NaN	80.000000	1.586300
75%	NaN	NaN	NaN	158.920000	3.166500
max	NaN	NaN	NaN	1166.000000	21.062900

```
[5]: # Check if there is null data
food_prices.isnull().sum()
```

```
[5]: date          0
      region        0
      city/province 0
      market        0
      latitude       0
      longitude      0
      category       0
      commodity      0
      unit           0
      priceflag      0
      pricetype      0
      currency       0
      price          0
      usdprice       0
      dtype: int64
```

```
[6]: # Check if there is duplication of data
food_prices.duplicated().sum()
```

```
[6]: 0
```

1.0.2 Convert column into correct data type

```
[56]: # Convert date column into datetime
food_prices['date'] = pd.to_datetime(food_prices['date'], dayfirst=False)
```

1.0.3 Augement data with additional columns

```
[8]: # Add Year and Month Column
food_prices['year'] = food_prices['date'].dt.year
food_prices['month'] = food_prices['date'].dt.month
```

1.0.4 Drop the unnecessary column

```
[9]: food_prices = food_prices.  
      ↪drop(food_prices[['latitude', 'longitude', 'currency', 'priceflag']], axis=1)
```

```
[10]: # Rename the column price into price_php  
food_prices.rename(columns={'price': 'price_php'}, inplace=True)
```

```
[11]: food_prices.head()
```

```
[11]:
```

	date	region	city/province	market	\
0	2000-01-15	National Capital region	Metropolitan Manila	Metro Manila	
1	2000-01-15	National Capital region	Metropolitan Manila	Metro Manila	
2	2000-01-15	National Capital region	Metropolitan Manila	Metro Manila	
3	2000-01-15	National Capital region	Metropolitan Manila	Metro Manila	
4	2000-01-15	National Capital region	Metropolitan Manila	Metro Manila	

	category	commodity	unit	pricetype	price_php	\
0	cereals and tubers	Maize flour (yellow)	KG	Retail	15.00	
1	cereals and tubers	Rice (milled, superior)	KG	Retail	20.00	
2	cereals and tubers	Rice (milled, superior)	KG	Wholesale	18.35	
3	cereals and tubers	Rice (regular, milled)	KG	Retail	18.00	
4	cereals and tubers	Rice (regular, milled)	KG	Wholesale	16.35	

	usdprice	year	month
0	0.3717	2000	1
1	0.4957	2000	1
2	0.4548	2000	1
3	0.4461	2000	1
4	0.4052	2000	1

1.0.5 1. What are the prices of foods based on their categories over the years?

```
[12]: # Extract the foods based on their category  
food_cereals = food_prices[food_prices['category'] == 'cereals and tubers']  
food_meats = food_prices[food_prices['category'] == 'meat, fish and eggs']  
food_nuts = food_prices[food_prices['category'] == 'pulses and nuts']  
food_veg = food_prices[food_prices['category'] == 'vegetables and fruits']
```

```
[13]: # Group it according to their years and get the average prices of it  
food_cereals_grp = food_cereals.groupby('year').mean(numeric_only = True)  
food_meats_grp = food_meats.groupby('year').mean(numeric_only = True)  
food_nuts_grp = food_nuts.groupby('year').mean(numeric_only = True)  
food_veg_grp = food_veg.groupby('year').mean(numeric_only = True)
```

```
[14]: # Create subplots  
fig, axs = plt.subplots(2, 2, figsize=(12, 8))  
colors = ['lightsteelblue', 'cornflowerblue', 'royalblue', 'navy']
```

```

# Plot each group on separate subplots
axs[0, 0].plot(food_cereals_grp.index, food_cereals_grp['price_php'],
    ↪marker='o', color=colors[0])
axs[0, 0].set_title('Cereals and Tubers')
axs[0, 0].set_xlabel('Year')
axs[0, 0].set_ylabel('Price in PHP')

axs[0, 1].plot(food_meats_grp.index, food_meats_grp['price_php'], marker='o',
    ↪color=colors[1])
axs[0, 1].set_title('Meats, Fish and Eggs')
axs[0, 1].set_xlabel('Year')
axs[0, 1].set_ylabel('Price in PHP')

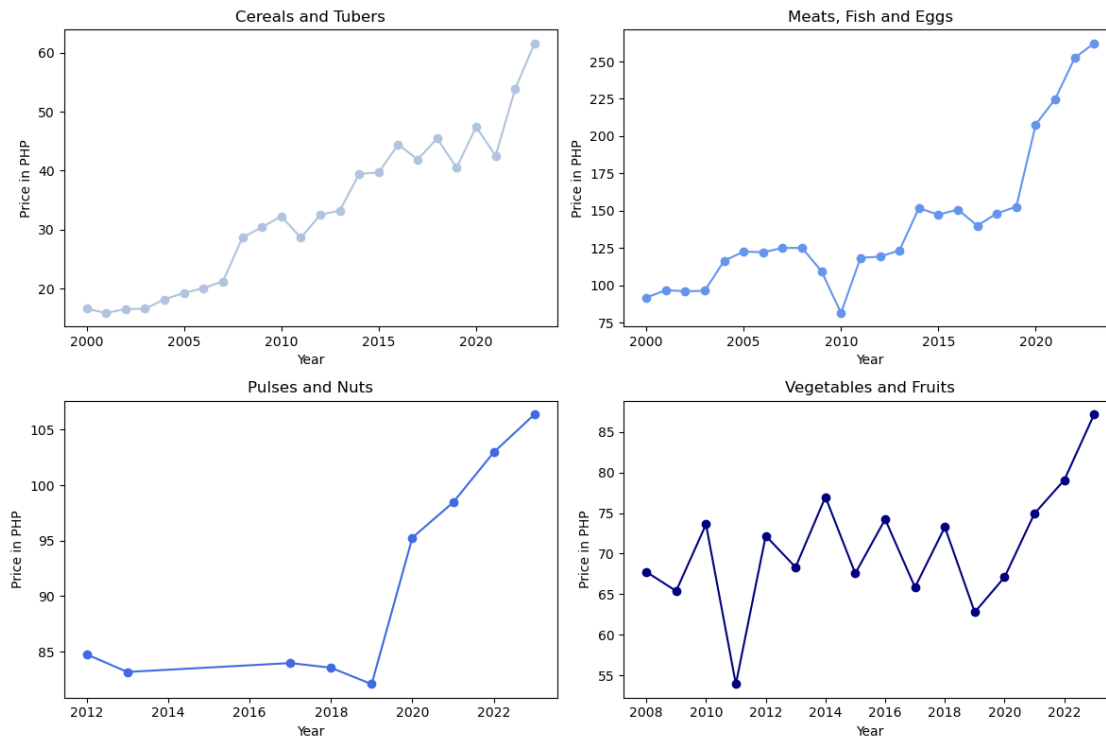
axs[1, 0].plot(food_nuts_grp.index, food_nuts_grp['price_php'], marker='o',
    ↪color=colors[2])
axs[1, 0].set_title('Pulses and Nuts')
axs[1, 0].set_xlabel('Year')
axs[1, 0].set_ylabel('Price in PHP')

axs[1, 1].plot(food_veg_grp.index, food_veg_grp['price_php'], marker='o',
    ↪color=colors[3])
axs[1, 1].set_title('Vegetables and Fruits')
axs[1, 1].set_xlabel('Year')
axs[1, 1].set_ylabel('Price in PHP')

# Adjust layout
plt.tight_layout()

# Display the plot
plt.show()

```



1.0.6 2. What are the average prices of different foods based on cereal and tuber commodities?

```
[15]: # Check its value in tabular format
food_cereals_commodity = food_cereals.groupby('commodity').mean(numeric_only =_
↪True)
food_cereals_commodity
```

```
[15]:
```

	price_php	usdprice	year	month
commodity				
Maize (white)	17.602381	0.365075	2012.584127	6.701587
Maize (yellow)	23.310014	0.455083	2017.526093	6.516220
Maize flour (white)	22.502729	0.476735	2009.375267	6.445629
Maize flour (yellow)	20.912271	0.441249	2009.353712	6.448326
Potatoes (Irish)	82.216344	1.621449	2017.794979	6.409780
Rice (milled, superior)	30.554382	0.644484	2009.447018	6.457544
Rice (paddy)	14.577440	0.328681	2009.855422	6.362952
Rice (premium)	45.453482	0.926445	2020.000000	8.473461
Rice (regular, milled)	34.111007	0.683148	2016.154658	6.483924
Rice (special)	50.802469	0.977202	2021.590666	6.295333
Rice (well milled)	42.761923	0.825025	2021.494423	6.429134
Semolina (white)	34.043766	0.665719	2021.062500	6.961538
Semolina (yellow)	29.489088	0.581289	2020.838235	7.302941

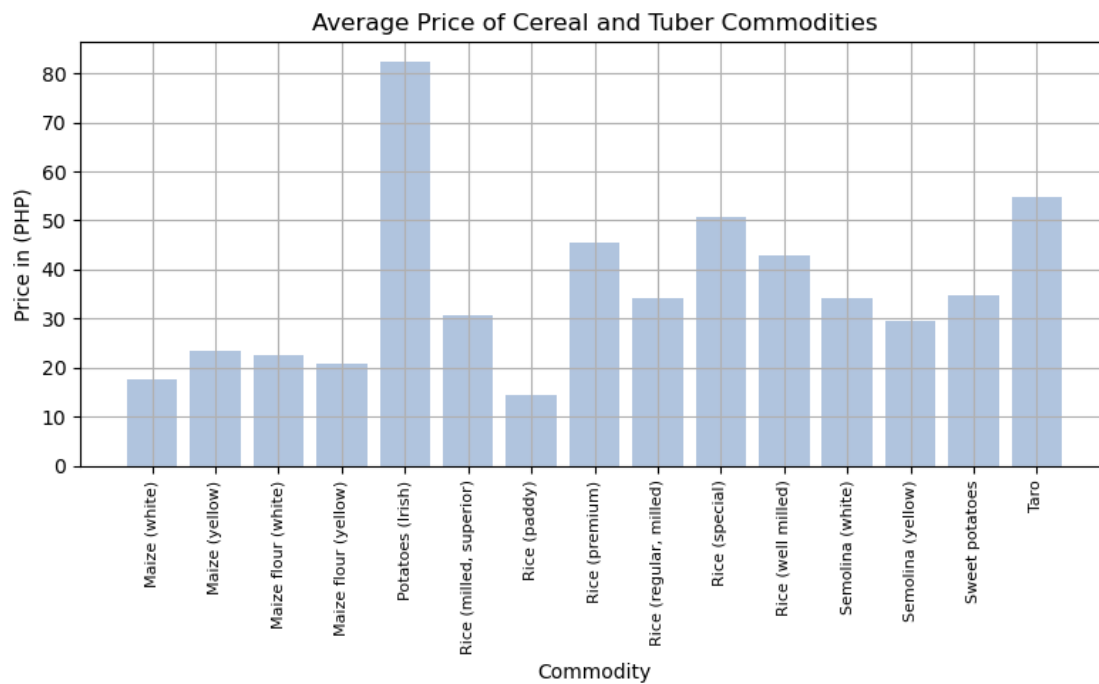
Sweet potatoes	34.854404	0.715550	2016.251445	6.748194
Taro	54.781201	1.100345	2020.569024	7.627385

```
[16]: # Values
cereals_group = food_cereals.groupby('commodity')
average_prices = cereals_group['price_php'].mean()

# Plotting
plt.figure(figsize=(8, 5))
plt.bar(average_prices.index, average_prices.values, color= colors[0])
plt.xticks(rotation='vertical', size=8)
plt.title('Average Price of Cereal and Tuber Commodities')
plt.xlabel('Commodity')
plt.ylabel('Price in (PHP)')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.7 3. What are the average prices of different foods based on meat, fish and eggs commodities?

```
[17]: # Check its value in tabular format
food_meats_commodity = food_meats.groupby('commodity').mean(numeric_only = True)
food_meats_commodity
```

```
[17]:
```

	price_php	usdprice	year \
commodity			
Anchovies	143.070315	2.760217	2021.483254
Chicken	130.802500	2.669101	2020.000000
Crab	318.336297	6.172280	2021.381669
Eggs	6.588257	0.131344	2018.529018
Eggs (duck)	8.895130	0.174999	2021.211460
Fish (fresh)	117.450311	2.485786	2014.060656
Fish (frigate tuna)	161.713752	3.219353	2020.517241
Fish (mackerel, fresh)	199.179325	3.916124	2020.708589
Fish (milkfish)	189.262851	3.645045	2021.473290
Fish (redbelly yellowtail fusilier)	261.012299	5.091756	2021.156561
Fish (roundscad)	185.614170	3.581907	2021.444405
Fish (slipmouth)	192.945194	3.785784	2021.144876
Fish (threadfin bream)	268.889487	5.282732	2020.910769
Fish (tilapia)	148.326045	2.859232	2021.476083
Meat (beef)	354.339791	6.853757	2021.386139
Meat (beef, chops with bones)	258.495050	5.122917	2018.820733
Meat (chicken, whole)	163.257027	3.271382	2018.354890
Meat (pork)	233.560284	4.636326	2016.856418
Meat (pork, hock)	195.083050	3.916536	2020.322302
Meat (pork, with bones)	260.820101	5.028031	2021.436502
Shrimp (endeavor)	398.397759	7.787631	2020.930425
Shrimp (tiger)	553.253425	10.684839	2021.505576

```
month
```

commodity	
Anchovies	6.418262
Chicken	8.588235
Crab	6.518924
Eggs	6.434440
Eggs (duck)	6.654843
Fish (fresh)	6.316393
Fish (frigate tuna)	7.772414
Fish (mackerel, fresh)	7.616564
Fish (milkfish)	6.450493
Fish (redbelly yellowtail fusilier)	6.857919
Fish (roundscad)	6.470795
Fish (slipmouth)	6.825972
Fish (threadfin bream)	7.236923
Fish (tilapia)	6.449341

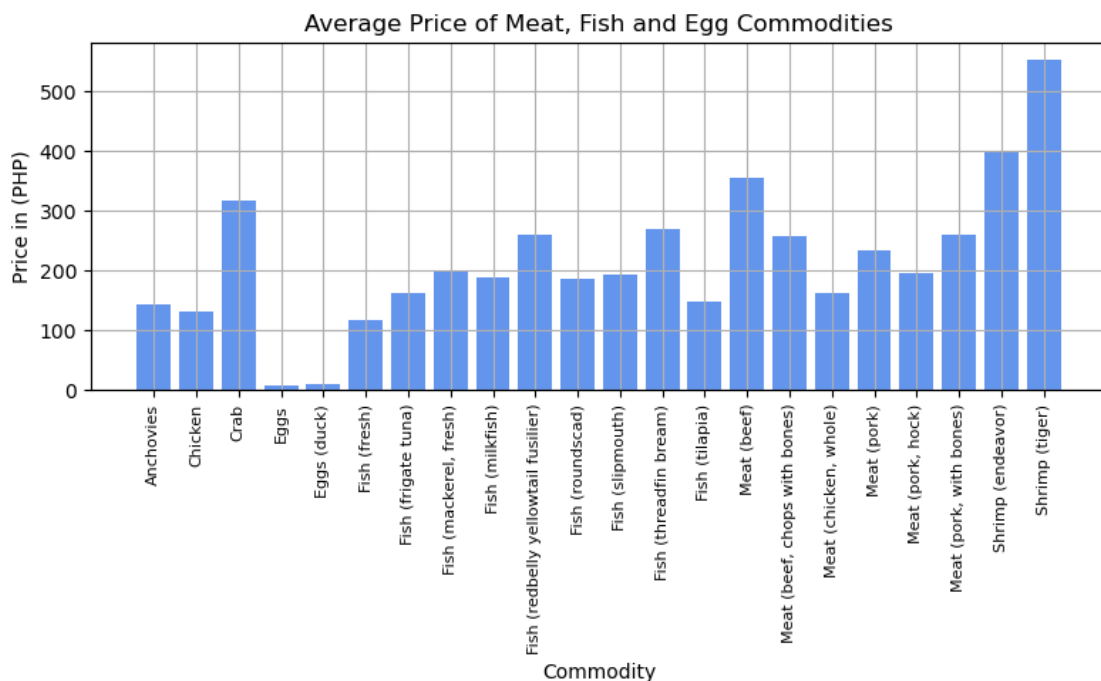
Meat (beef)	6.550647
Meat (beef, chops with bones)	6.386827
Meat (chicken, whole)	6.447243
Meat (pork)	6.420598
Meat (pork, hock)	8.043165
Meat (pork, with bones)	6.487162
Shrimp (endeavor)	7.344340
Shrimp (tiger)	6.380112

```
[18]: # Values
meats_group = food_meats.groupby('commodity')
average_prices = meats_group['price_php'].mean()

# Plotting
plt.figure(figsize=(8, 5))
plt.bar(average_prices.index, average_prices.values, color= colors[1])
plt.xticks(rotation='vertical', size=8)
plt.title('Average Price of Meat, Fish and Egg Commodities')
plt.xlabel('Commodity')
plt.ylabel('Price in (PHP)')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.8 4. What are the average prices of different foods based on pulse and nut commodities?

```
[19]: # Check its value in tabular format
food_nuts_commodity = food_nuts.groupby('commodity').mean(numeric_only = True)
food_nuts_commodity
```

```
[19]:
```

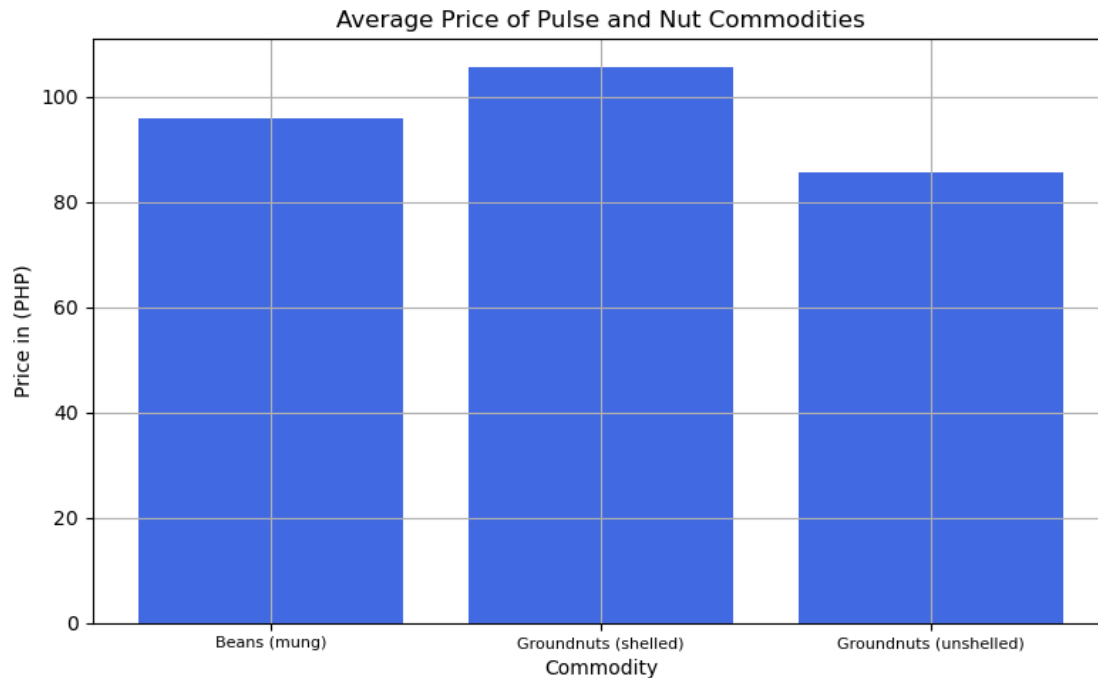
	price_php	usdprice	year	month
commodity				
Beans (mung)	95.892711	1.882528	2020.310321	6.573266
Groundnuts (shelled)	105.656252	2.055130	2021.303342	6.655013
Groundnuts (unshelled)	85.606886	1.756495	2019.329829	6.725361

```
[20]: # Values
nuts_group = food_nuts.groupby('commodity')
average_prices = nuts_group['price_php'].mean()

# Plotting
plt.figure(figsize=(8, 5))
plt.bar(average_prices.index, average_prices.values, color= colors[2])
plt.xticks(size=8)
plt.title('Average Price of Pulse and Nut Commodities')
plt.xlabel('Commodity')
plt.ylabel('Price in (PHP)')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.9 5. What are the average prices of different foods based on vegetable and fruit commodities?

```
[21]: # Check its value in tabular format
food_veg_commodity = food_veg.groupby('commodity').mean(numeric_only = True)
food_veg_commodity
```

```
[21]:
```

commodity	price_php	usdprice	year	month
Bananas (lakatan)	67.554881	1.307257	2021.450020	6.512343
Bananas (latundan)	45.934332	0.891649	2021.352273	6.648674
Bananas (saba)	36.200947	0.710288	2021.015909	7.120455
Beans (green, fresh)	90.223809	1.759541	2021.164058	6.860290
Beans (string)	65.528576	1.273776	2021.233986	6.773425
Bitter melon	88.812625	1.713834	2021.453802	6.467780
Bottle gourd	35.528016	0.701354	2021.061053	6.882632
Cabbage	66.813808	1.328457	2018.378677	6.432714
Cabbage (chinese)	74.526775	1.453204	2021.124661	6.963415
Calamansi	71.227346	1.371292	2021.330853	6.687004
Carrots	81.285737	1.614221	2018.375282	6.400902
Choko	37.254118	0.724328	2021.287304	6.650041
Coconut	24.782074	0.479415	2021.432712	6.492409
Eggplants	71.049634	1.366791	2021.480732	6.445787
Garlic	169.452478	3.456442	2018.054362	6.266330
Ginger	118.710807	2.325668	2021.354541	6.627478

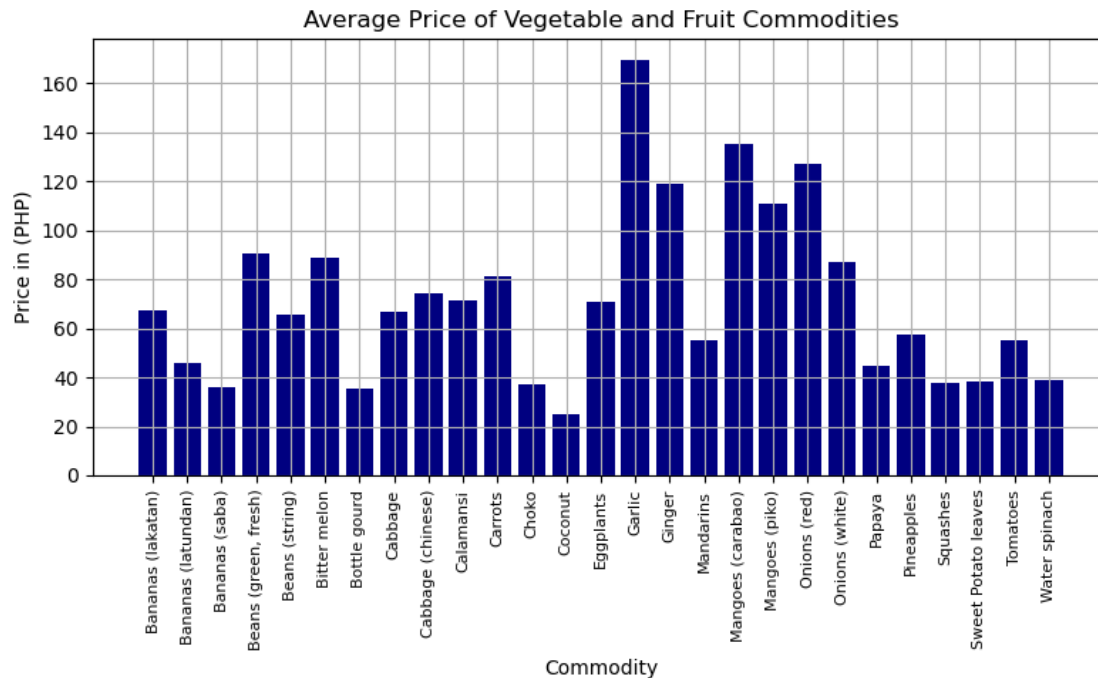
Mandarins	55.246019	1.088677	2021.245370	6.731481
Mangoes (carabao)	135.348905	2.624672	2021.352941	6.582449
Mangoes (piko)	110.994143	2.174957	2020.956175	6.848606
Onions (red)	126.908194	2.506834	2018.080074	6.465486
Onions (white)	86.750117	1.795826	2016.180712	6.802346
Papaya	44.405408	0.897056	2020.304000	8.132800
Pineapples	57.184409	1.128158	2020.845304	7.217680
Squashes	37.580402	0.726678	2021.380880	6.682218
Sweet Potato leaves	38.031376	0.756837	2020.603468	7.678613
Tomatoes	55.348039	1.099578	2018.325331	6.412614
Water spinach	38.885556	0.774976	2020.581633	7.692744

```
[22]: # Values
veg_group = food_veg.groupby('commodity')
average_prices = veg_group['price_php'].mean()

# Plotting
plt.figure(figsize=(8, 5))
plt.bar(average_prices.index, average_prices.values, color= colors[3])
plt.xticks(rotation='vertical', size=8)
plt.title('Average Price of Vegetable and Fruit Commodities')
plt.xlabel('Commodity')
plt.ylabel('Price in (PHP)')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.10 6. What are the prices of Rice (milled,superior) and Rice(milled, regular) over the available years?

```
[23]: # Extract the Rice (milled, superior) from the dataset
rice_superior = food_prices[food_prices['commodity'] == 'Rice (milled,
↪superior)']
rice_superior = rice_superior.groupby('year').mean(numeric_only = True)
rice_superior
```

```
[23]:
```

	price_php	usdprice	month
year			
2000	18.746528	0.425686	6.500000
2001	18.475000	0.362428	6.500000
2002	19.415070	0.376337	6.549296
2003	19.443403	0.359081	6.500000
2004	20.315211	0.363980	6.563380
2005	21.979306	0.400479	6.500000
2006	22.475625	0.439017	6.500000
2007	23.843542	0.518858	6.500000
2008	31.388056	0.708152	6.500000
2009	32.684028	0.688494	6.500000
2010	32.586111	0.724411	6.500000
2011	33.084722	0.763632	6.500000
2012	33.992431	0.805844	6.500000

2013	35.766806	0.844035	6.500000
2014	41.254514	0.934321	6.500000
2015	40.826087	0.899724	6.391304
2016	40.620709	0.859824	6.517730
2017	41.307483	0.820481	6.482517
2018	44.389008	0.842687	6.442748
2019	40.793697	0.787555	6.470588
2020	38.031818	0.749909	1.454545

```
[24]: # Extract the Rice (regular, milled) from the dataset
rice_regular = food_prices[food_prices['commodity'] == 'Rice (regular, milled)']
rice_regular = rice_regular.groupby('year').mean(numeric_only = True)
rice_regular
```

```
[24]:
```

	price_php	usdprice	month
year			
2000	17.152869	0.388108	6.680328
2001	16.551230	0.324330	6.696721
2002	17.312353	0.335582	6.655462
2003	17.283984	0.319020	6.650407
2004	18.438049	0.330301	6.634146
2005	20.309024	0.369985	6.674797
2006	20.657280	0.403727	6.640000
2007	22.212742	0.484293	6.661290
2008	28.654956	0.646273	6.654867
2009	29.604602	0.623711	6.654867
2010	30.369454	0.675630	6.584699
2011	30.665398	0.707819	6.654867
2012	31.907514	0.756701	6.594595
2013	33.031226	0.782694	6.012903
2014	38.674899	0.876980	7.342282
2015	36.864205	0.811941	6.505682
2016	36.373757	0.769623	6.574586
2017	36.950323	0.733184	6.875576
2018	40.755750	0.772930	6.600000
2019	36.455789	0.701989	5.368421
2020	37.362736	0.760898	8.300166
2021	37.577252	0.764284	6.510571
2022	39.088476	0.719223	6.500000
2023	40.309948	0.728883	4.000000

```
[25]: # Extract the Rice (special) from the dataset
rice_paddy = food_prices[food_prices['commodity'] == 'Rice (paddy)']
rice_paddy = rice_paddy.groupby('year').mean(numeric_only = True)
rice_paddy
```

```
[25]:      price_php  usdprice      month
year
2008  13.714146  0.310783  6.548780
2009  14.262883  0.300466  6.582822
2010  14.516981  0.322806  6.559748
2012  15.814101  0.375481  6.870504
2013  15.361282  0.379018  2.051282
```

```
[26]: # Extract the Rice (special) from the dataset
rice_special = food_prices[food_prices['commodity'] == 'Rice (special)']
rice_special = rice_special.groupby('year').mean(numeric_only = True)
rice_special
```

```
[26]:      price_php  usdprice      month
year
2020  50.889495  1.037444  8.515152
2021  49.886670  1.014688  6.510989
2022  50.918259  0.937073  6.500000
2023  52.048662  0.941171  4.000000
```

```
[27]: # Extract the Rice (well milled) from the dataset
rice_wmilled = food_prices[food_prices['commodity'] == 'Rice (well milled)']
rice_wmilled = rice_wmilled.groupby('year').mean(numeric_only = True)
rice_wmilled
```

```
[27]:      price_php  usdprice      month
year
2020  42.109800  0.858351  8.489149
2021  42.063011  0.855511  6.510989
2022  43.086183  0.792922  6.500000
2023  44.016684  0.795924  4.000000
```

```
[28]: # X value from 2000 - 2023
years = range(2000,2024)

# Create and Display the plot in line chart
fig, ax = plt.subplots(figsize=(12, 6))
ax.plot(rice_superior.index, rice_superior['price_php'], label='Rice (milled, superior)',marker='.',color = colors[0])
ax.plot(rice_regular.index, rice_regular['price_php'], label='Rice (regular, milled)',marker='.',color = colors[1])
ax.plot(rice_special.index, rice_special['price_php'], label='Rice (special)',marker='.',color = colors[2])
ax.plot(rice_wmilled.index, rice_wmilled['price_php'], label='Rice (well milled)',marker='.',color = colors[3])
ax.plot(rice_paddy.index, rice_paddy['price_php'], label='Rice (paddy)',marker='.')
```

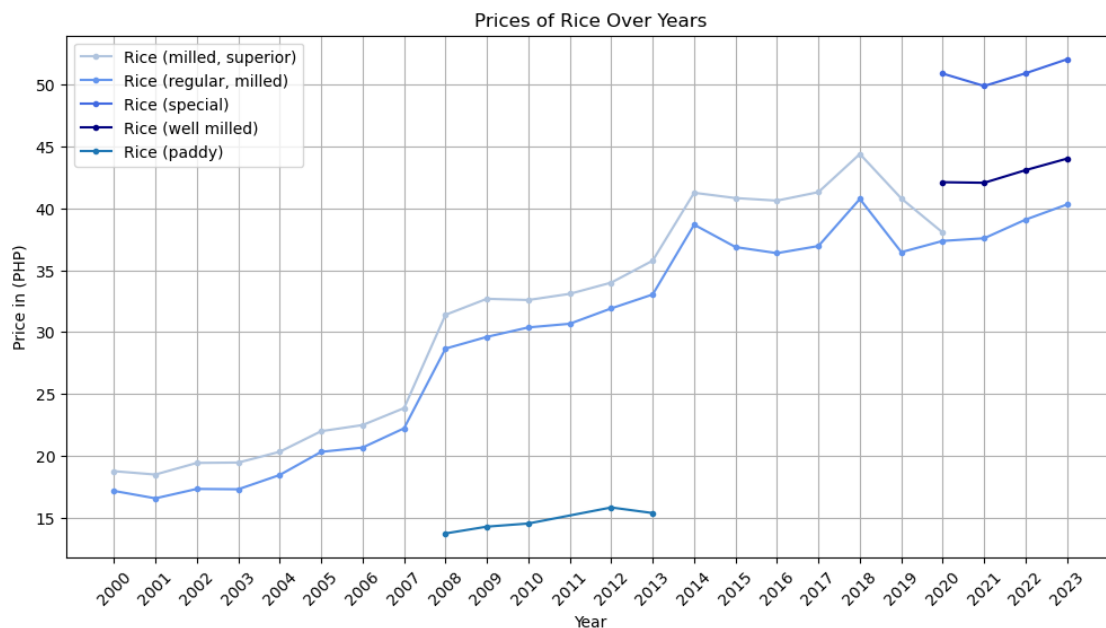
```

# Set labels and title
ax.set_xlabel('Year')
ax.set_ylabel('Price in (PHP)')
ax.set_title('Prices of Rice Over Years')
plt.xticks(years, rotation=45)

# Add a legend
ax.legend()

# Display the plot
plt.grid(True)
plt.show()

```



1.0.11 7. What are the prices of Meat(beef), Meat(pork), Meat(chicken, whole) over the years?

```

[29]: # Extract the Meat (beef, chops with bones) from the dataset
beef = food_prices[food_prices['commodity'] == 'Meat (beef, chops with bones)']
beef = beef.groupby('year').mean(numeric_only = True)
beef

```

```

[29]:      price_php  usdprice  month
year
2008  182.587500   4.131308   6.500000
2009  166.799531   3.514164   6.500000

```


2010	202.575833	4.505183	6.500000
2011	179.705260	4.147907	6.500000
2012	181.058087	4.280183	5.934426
2013	181.822805	4.418485	3.329268
2014	186.031111	4.224127	9.765432
2015	186.914857	4.128223	6.068571
2016	192.299737	4.070317	6.521053
2017	203.858895	4.047926	6.531579
2018	225.726919	4.286261	6.474747
2019	233.946988	4.483959	3.024096
2020	244.354164	4.984138	8.520067
2021	284.136624	5.776902	6.511601
2022	319.971470	5.883811	6.500000
2023	333.674504	6.033685	4.000000

```
[30]: # Extract the Meat (pork) from the dataset
pork = food_prices[food_prices['commodity'] == 'Meat (pork)']
pork = pork.groupby('year').mean(numeric_only = True)
pork
```

```
[30]:      price_php  usdprice      month
year
2008  182.587500   4.131308   6.500000
2009  166.799531   3.514164   6.500000
2010  202.575833   4.505183   6.500000
2011  179.705260   4.147907   6.500000
2012  181.058087   4.280183   5.934426
2013  181.822805   4.418485   3.329268
2014  186.031111   4.224127   9.765432
2015  186.914857   4.128223   6.068571
2016  192.299737   4.070317   6.521053
2017  203.858895   4.047926   6.531579
2018  225.726919   4.286261   6.474747
2019  233.946988   4.483959   3.024096
2020  244.354164   4.984138   8.520067
2021  284.136624   5.776902   6.511601
2022  319.971470   5.883811   6.500000
2023  333.674504   6.033685   4.000000
```

```
[31]: # Extract the Meat (chicken, whole) from the dataset
chicken = food_prices[food_prices['commodity'] == 'Meat (chicken, whole)']
chicken = chicken.groupby('year').mean(numeric_only = True)
chicken
```

```
[31]:      price_php  usdprice      month
year
2008  117.743333   2.664225   6.500000
```

2009	128.041615	2.698476	6.500000
2010	128.275276	2.851928	6.607362
2011	128.417917	2.964054	6.500000
2012	128.355568	3.034390	5.945946
2013	131.351235	3.186914	3.456790
2014	140.588765	3.192156	9.765432
2015	138.006492	3.039098	6.481675
2016	140.535579	2.974685	6.521053
2017	144.518550	2.870191	6.580000
2018	152.879698	2.903512	6.472362
2019	148.120357	2.838644	3.023810
2020	154.602669	3.153125	8.518946
2021	176.874398	3.600055	6.512285
2022	193.885586	3.563077	6.500000
2023	202.199085	3.656467	4.000000

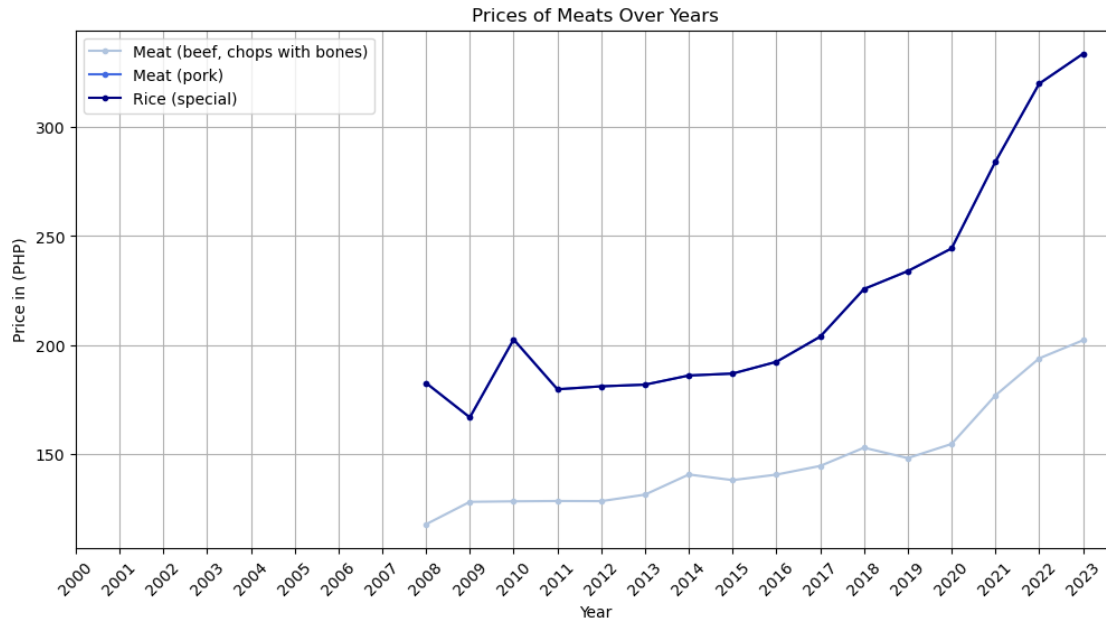
```
[32]: # X value from 2000 - 2023
years = range(2000,2024)

# Create and Display the plot lin line chart
fig, ax = plt.subplots(figsize=(12, 6))
ax.plot(chicken.index, chicken['price_php'], label= 'Meat (beef, chops with_
↪bones)',marker='.',color = colors[0])
ax.plot(beef.index, beef['price_php'], label= 'Meat (pork)',marker='.',color =_
↪colors[2])
ax.plot(pork.index, pork['price_php'], label='Rice (special)', marker='.',color_
↪= colors[3])

# Set labels and title
ax.set_xlabel('Year')
ax.set_ylabel('Price in (PHP)')
ax.set_title('Prices of Meats Over Years')
plt.xticks(years, rotation=45)

# Add a legend
ax.legend()

# Display the plot
plt.grid(True)
plt.show()
```



1.0.12 8. What are the prices of Beans (mung), Groundnuts (shelled), and Groundnuts (unshelled) over the years?

```
[33]: # Extract the Beans (mung) from the dataset
mung = food_prices[food_prices['commodity'] == 'Beans (mung)']
mung = mung.groupby('year').mean(numeric_only = True)
mung
```

```
[33]:      price_php  usdprice  month
year
2012    76.064153    1.818100    8.457627
2013    72.879667    1.790422    2.500000
2017    83.968346    1.660407    8.526316
2018    83.548090    1.587099    6.472362
2019    82.069643    1.572851    3.023810
2020    94.301675    1.922791    8.487562
2021    99.496284    2.023396    6.500000
2022   102.471574    1.886036    6.500000
2023   103.783413    1.876724    4.000000
```

```
[34]: # Extract the Groundnuts (shelled) from the dataset
shelled = food_prices[food_prices['commodity'] == 'Groundnuts (shelled)']
shelled = shelled.groupby('year').mean(numeric_only = True)
shelled
```

```
[34]:      price_php  usdprice      month
year
2020  103.445852   2.109492   8.496212
2021  101.608100   2.066870   6.500000
2022  107.912287   1.984959   6.500000
2023  113.735548   2.056660   4.000000
```

```
[35]: # Extract the Groundnuts (unshelled) from the dataset
unshelled = food_prices[food_prices['commodity'] == 'Groundnuts (unshelled)']
unshelled = unshelled.groupby('year').mean(numeric_only = True)
unshelled
```

```
[35]:      price_php  usdprice      month
year
2012  94.336636   2.254625   8.364486
2013  94.810377   2.328892   2.528302
2020  78.466622   1.597845   8.346847
2021  78.213636   1.590662   6.500000
2022  88.659423   1.631251   6.500000
2023  92.892418   1.679697   4.000000
```

```
[36]: # X value from 2012 - 2023
years = range(2012,2024)

# Create and Display the plot in line chart
fig, ax = plt.subplots(figsize=(12, 6))
ax.plot(mung.index, mung['price_php'], label= 'Beans (mung)',marker='o',color = colors[0])

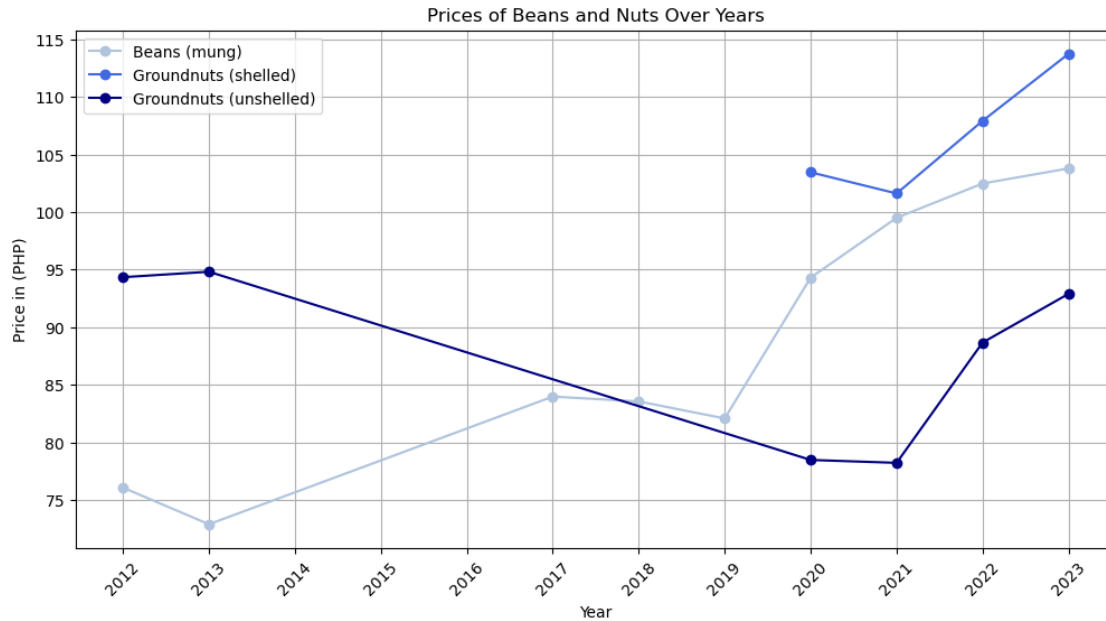
ax.plot(shelled.index, shelled['price_php'], label= 'Groundnuts (shelled)',marker='o',color = colors[2])

ax.plot(unshelled.index, unshelled['price_php'], label='Groundnuts (unshelled)', marker='o',color = colors[3])

# Set labels and title
ax.set_xlabel('Year')
ax.set_ylabel('Price in (PHP)')
ax.set_title('Prices of Beans and Nuts Over Years')
plt.xticks(years, rotation=45)

# Add a legend
ax.legend()

# Display the plot
plt.grid(True)
plt.show()
```



1.0.13 9. What are the prices of Onions (red) and Onions (white) over the years?

```
[37]: # Extract the Onions (red) from the dataset
red = food_prices[food_prices['commodity'] == 'Onions (red)']
red = red.groupby('year').mean(numeric_only = True)
red
```

```
[37]:
```

	price_php	usdprice	month
year			
2008	82.767839	1.871806	6.608040
2009	67.103529	1.415738	6.500000
2010	58.789902	1.306737	6.500000
2012	82.263114	1.948488	6.161677
2013	71.406190	1.735879	3.357143
2014	65.050000	1.476048	9.835443
2015	69.633403	1.526684	6.481675
2016	97.526684	2.064194	6.521053
2017	81.257000	1.612912	6.580000
2018	94.127136	1.788502	6.472362
2019	74.984643	1.436796	3.023810
2020	139.471250	2.852604	8.513158
2021	130.180913	2.646285	6.512853
2022	160.615765	2.941762	6.500000
2023	252.350047	4.565492	4.000000

```
[38]: # Extract the Onions (white) from the dataset
white = food_prices[food_prices['commodity'] == 'Onions (white)']
white = white.groupby('year').mean(numeric_only = True)
white
```

```
[38]:      price_php  usdprice      month
year
2008    69.424534    1.569771    6.503106
2009    62.652532    1.321978    6.664557
2010    56.782788    1.265775    6.581818
2012    70.799448    1.675895    6.006897
2013    66.500270    1.615334    3.351351
2014    73.814833    1.676572    9.566667
2015    67.112632    1.473055    6.548872
2016    82.689462    1.751570    6.530769
2017    75.627986    1.499781    6.597122
2018    81.330500    1.543430    6.471429
2019    69.072973    1.323678    3.027027
2020   101.835925    2.079624    8.508079
2021   105.012183    2.136987    6.500000
2022   156.622639    2.831719    6.500000
2023   202.125000    3.655286    4.000000
```

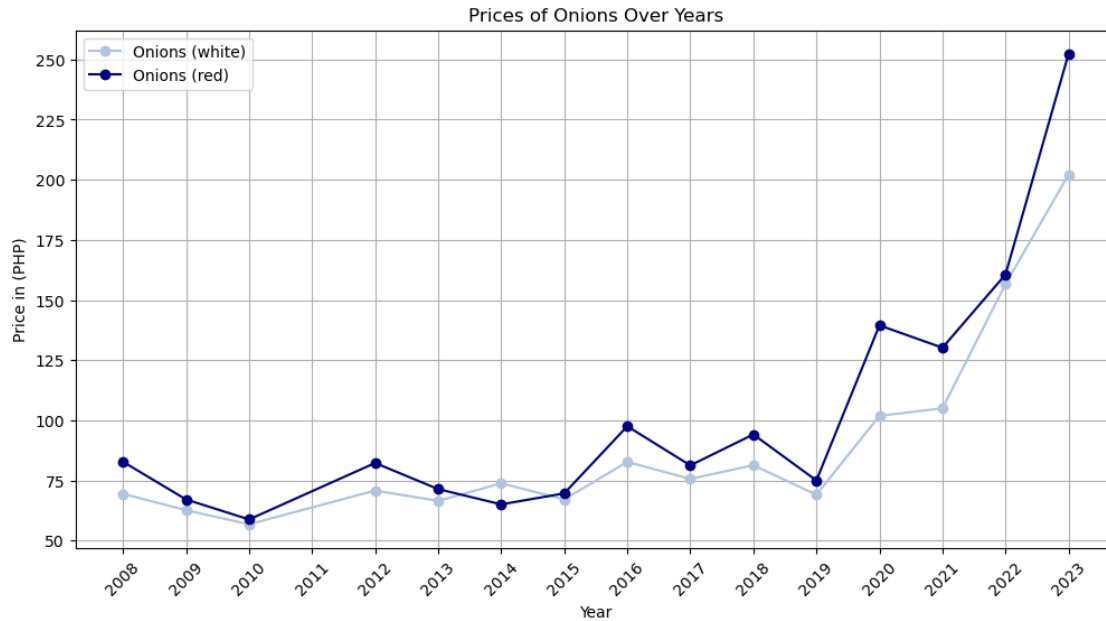
```
[39]: # X value from 2008 - 2023
years = range(2008,2024)

# Create and Display the plot in line chart
fig, ax = plt.subplots(figsize=(12, 6))
ax.plot(white.index, white['price_php'], label= 'Onions_
↳(white)',marker='o',color = colors[0])
ax.plot(red.index, red['price_php'], label= 'Onions (red)',marker='o',color =_
↳colors[3])

# Set labels and title
ax.set_xlabel('Year')
ax.set_ylabel('Price in (PHP)')
ax.set_title('Prices of Onions Over Years')
plt.xticks(years, rotation=45)

# Add a legend
ax.legend()

# Display the plot
plt.grid(True)
plt.show()
```



1.0.14 10. What are the average price of cereals and tubers per region?

```
[40]: # Extract the foods based on their category
food_cereals = food_prices[food_prices['category'] == 'cereals and tubers']
food_meats = food_prices[food_prices['category'] == 'meat, fish and eggs']
food_nuts = food_prices[food_prices['category'] == 'pulses and nuts']
food_veg = food_prices[food_prices['category'] == 'vegetables and fruits']
```

```
[41]: # Group by Region
reg_cereals = food_cereals.groupby('region').mean(numeric_only = True)
reg_cereals
```

```
[41]:
```

	price_php	usdprice	year	\
region				
Autonomous region in Muslim Mindanao	52.142586	1.030300	2018.975732	
Cordillera Administrative region	47.764341	0.940433	2019.896739	
National Capital region	33.078471	0.684074	2011.175381	
Region I	44.431037	0.880022	2018.863372	
Region II	38.414020	0.758985	2019.290958	
Region III	43.782958	0.872051	2017.108964	
Region IV-A	49.672782	0.983454	2019.652439	
Region IV-B	50.533045	0.994282	2019.539267	
Region IX	42.296324	0.835698	2017.796157	
Region V	50.201824	0.982982	2019.782866	
Region VI	42.941176	0.858614	2015.969369	
Region VII	37.547868	0.761098	2013.594051	

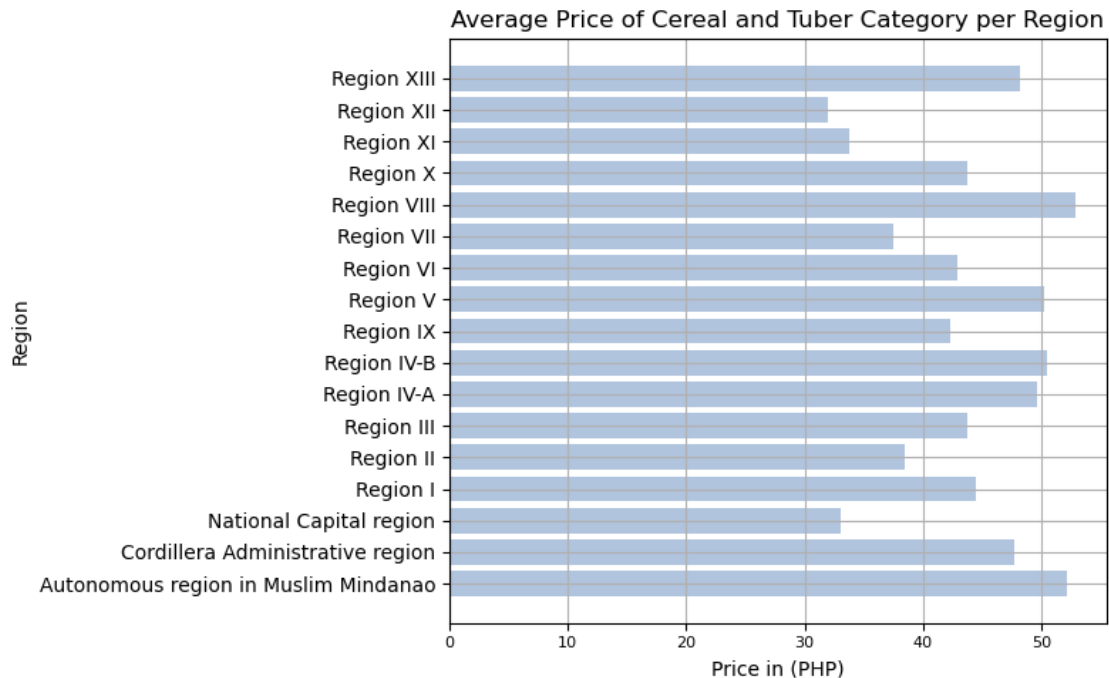
Region VIII	52.895866	1.039900	2019.476761
Region X	43.746881	0.864336	2019.142753
Region XI	33.741377	0.684489	2013.314845
Region XII	31.953721	0.649065	2013.350021
Region XIII	48.225120	0.953937	2019.229215

	month
region	
Autonomous region in Muslim Mindanao	6.499582
Cordillera Administrative region	6.519701
National Capital region	6.426208
Region I	6.650194
Region II	6.544315
Region III	6.597424
Region IV-A	6.637195
Region IV-B	6.666667
Region IX	6.521303
Region V	6.596750
Region VI	6.762613
Region VII	6.514872
Region VIII	6.582394
Region X	6.630307
Region XI	6.523980
Region XII	6.558198
Region XIII	6.532246

```
[42]: # Plotting
plt.figure(figsize=(8, 5))
plt.barh(reg_cereals.index, reg_cereals['price_php'], color= colors[0])
plt.xticks(size=8)
plt.title('Average Price of Cereal and Tuber Category per Region')
plt.xlabel('Price in (PHP)')
plt.ylabel('Region')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```

```
[43]: food_prices['category'].unique()
```

```
[43]: array(['cereals and tubers', 'meat, fish and eggs',
        'vegetables and fruits', 'pulses and nuts'], dtype=object)
```

1.0.15 11. What are the average price of meat, fish and eggs per region?

```
[44]: # Group by Region
reg_meats = food_meats.groupby('region').mean(numeric_only = True)
reg_meats
```

```
[44]:
```

	price_php	usdprice	year	\
region				
Autonomous region in Muslim Mindanao	165.968492	3.292692	2019.686805	
Cordillera Administrative region	219.801779	4.314243	2020.086177	
National Capital region	223.438706	4.410858	2016.643617	
Region I	210.335606	4.130096	2019.770362	
Region II	215.990342	4.240193	2019.827483	
Region III	241.496093	4.700136	2020.118820	
Region IV-A	249.102604	4.874288	2020.111526	
Region IV-B	213.209391	4.168453	2020.253727	
Region IX	199.896201	3.892709	2019.857804	
Region V	219.124264	4.296793	2020.164779	
Region VI	202.529022	3.946357	2019.864920	
Region VII	194.895731	3.858347	2018.612301	

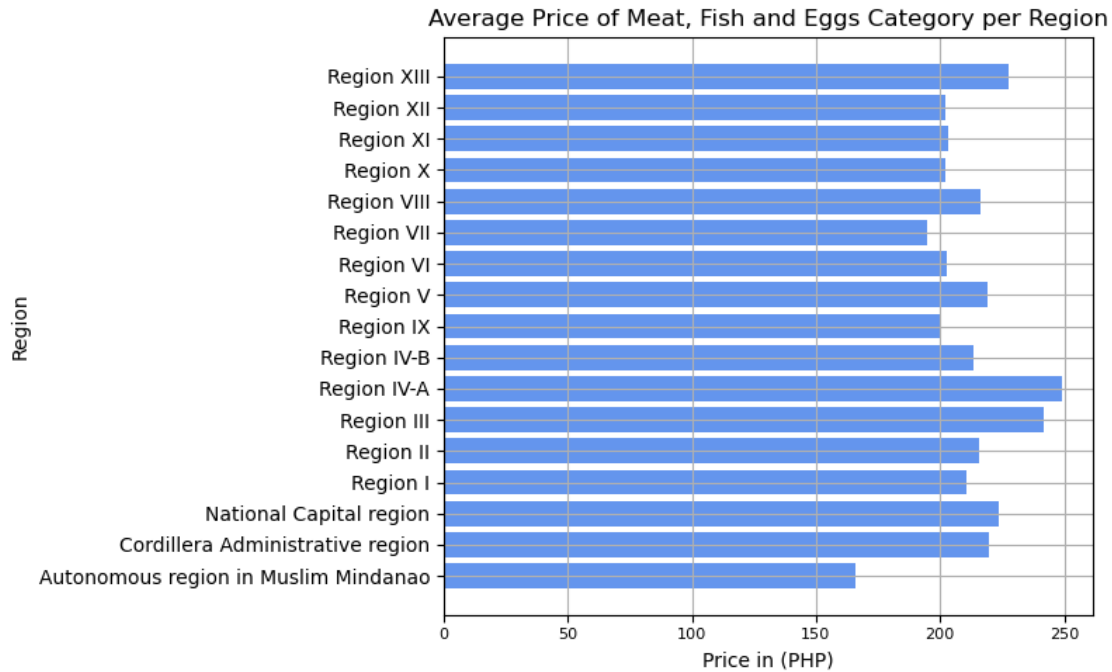
Region VIII	216.491233	4.228475	2020.335129
Region X	201.993398	3.962428	2019.931435
Region XI	202.983528	3.959116	2019.245335
Region XII	202.269130	3.967594	2018.999580
Region XIII	227.578172	4.439933	2020.181296

	month
region	
Autonomous region in Muslim Mindanao	6.632189
Cordillera Administrative region	6.563943
National Capital region	6.354610
Region I	6.567739
Region II	6.476884
Region III	6.516479
Region IV-A	6.600000
Region IV-B	6.534161
Region IX	6.517459
Region V	6.632300
Region VI	6.525681
Region VII	6.609112
Region VIII	6.586901
Region X	6.625703
Region XI	6.568417
Region XII	6.521227
Region XIII	6.534304

```
[45]: # Plotting
plt.figure(figsize=(8, 5))
plt.barh(reg_meats.index, reg_meats['price_php'], color= colors[1])
plt.xticks(size=8)
plt.title('Average Price of Meat, Fish and Eggs Category per Region')
plt.xlabel('Price in (PHP)')
plt.ylabel('Region')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.16 12. What are the average price of pulses and nuts per region?

```
[46]: # Group by Region
reg_nuts = food_nuts.groupby('region').mean(numeric_only = True)
reg_nuts
```

```
[46]:
```

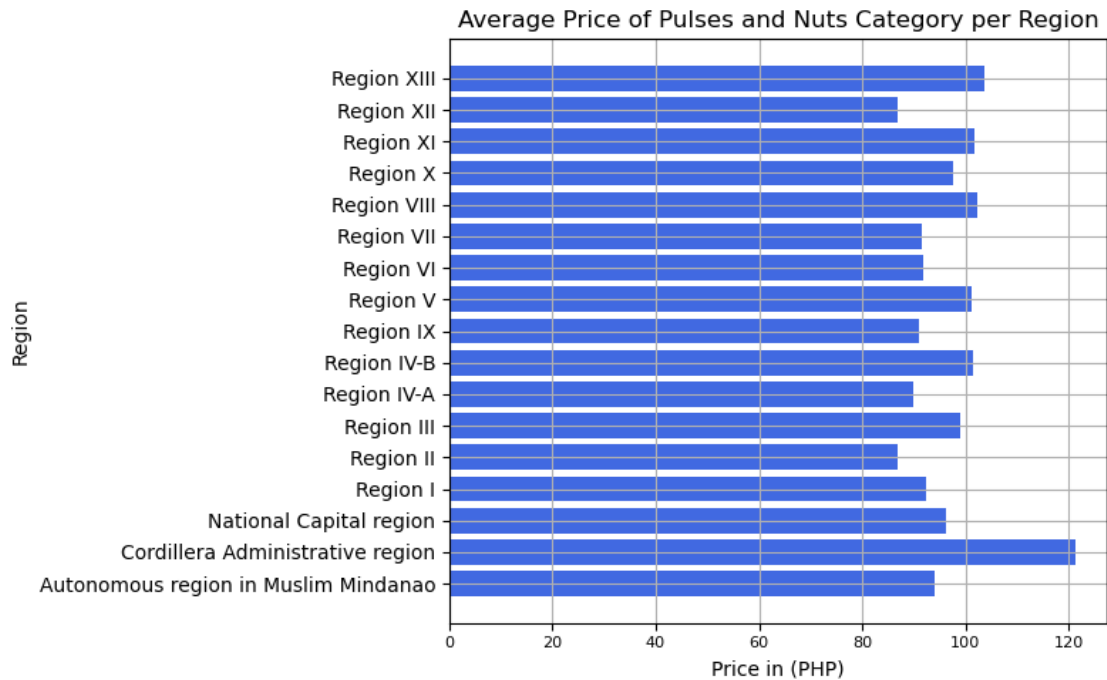
	price_php	usdprice	year \
region			
Autonomous region in Muslim Mindanao	94.111208	1.824522	2020.771812
Cordillera Administrative region	121.336272	2.410127	2020.346154
National Capital region	96.374444	1.933576	2019.088889
Region I	92.277158	1.830992	2020.246575
Region II	86.974496	1.712061	2020.596730
Region III	99.008195	1.938123	2020.943205
Region IV-A	89.877202	1.798361	2019.697248
Region IV-B	101.619321	1.974994	2021.000000
Region IX	90.886507	1.783762	2020.514706
Region V	101.297960	1.972852	2020.956284
Region VI	91.945664	1.859257	2019.517699
Region VII	91.557914	1.802797	2020.582781
Region VIII	102.226060	1.998560	2020.812000
Region X	97.520246	1.926580	2020.252632
Region XI	101.836838	1.989803	2020.654412
Region XII	86.803645	1.706634	2020.404682
Region XIII	103.609643	2.064267	2020.011905

region	month
Autonomous region in Muslim Mindanao	6.557047
Cordillera Administrative region	6.713018
National Capital region	6.585185
Region I	6.633562
Region II	6.482289
Region III	6.651116
Region IV-A	6.894495
Region IV-B	6.501171
Region IX	6.761029
Region V	6.398907
Region VI	7.247788
Region VII	6.665563
Region VIII	6.536000
Region X	6.666667
Region XI	6.563725
Region XII	6.548495
Region XIII	6.718254

```
[47]: # Plotting
plt.figure(figsize=(8, 5))
plt.barh(reg_nuts.index, reg_nuts['price_php'], color= colors[2])
plt.xticks(size=8)
plt.title('Average Price of Pulses and Nuts Category per Region')
plt.xlabel('Price in (PHP)')
plt.ylabel('Region')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.17 13. What are the average price of vegetables and fruits per region?

```
[48]: # Group by Region
reg_veg = food_veg.groupby('region').mean(numeric_only = True)
reg_veg
```

```
[48]:
```

region	price_php	usdprice	year	\
Autonomous region in Muslim Mindanao	70.741863	1.390529	2019.845730	
Cordillera Administrative region	76.795134	1.517722	2020.146930	
National Capital region	87.531794	1.739244	2017.904973	
Region I	70.414719	1.393061	2019.706256	
Region II	64.591358	1.288261	2019.608331	
Region III	75.564632	1.474074	2020.495916	
Region IV-A	81.430604	1.615207	2019.828786	
Region IV-B	85.484509	1.692565	2020.064854	
Region IX	65.232967	1.275358	2019.991204	
Region V	82.787676	1.627502	2020.266034	
Region VI	81.280037	1.601860	2020.069282	
Region VII	79.636059	1.577670	2019.731325	
Region VIII	82.734235	1.622589	2020.310377	
Region X	64.788797	1.276744	2020.165195	
Region XI	62.852949	1.238168	2019.981588	
Region XII	58.673521	1.161302	2019.927937	

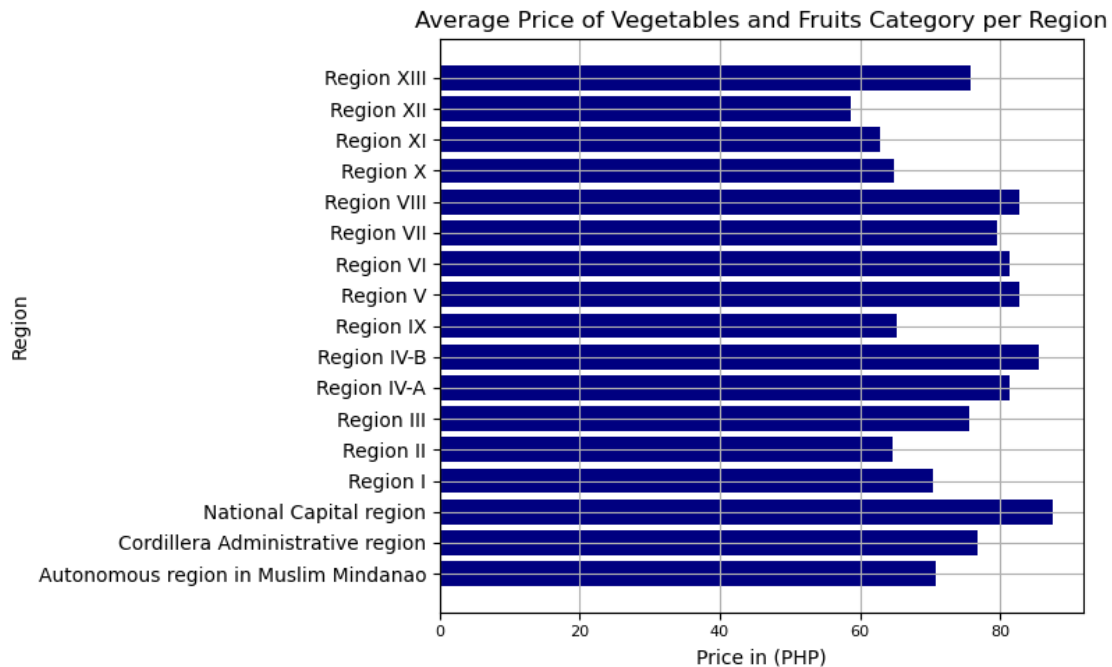
Region XIII	75.840019	1.492568	2020.286447
-------------	-----------	----------	-------------

	month
region	
Autonomous region in Muslim Mindanao	6.661846
Cordillera Administrative region	6.648246
National Capital region	6.373002
Region I	6.658304
Region II	6.642213
Region III	6.649164
Region IV-A	6.721882
Region IV-B	6.720193
Region IX	6.527740
Region V	6.581769
Region VI	6.739508
Region VII	6.607439
Region VIII	6.657765
Region X	6.620429
Region XI	6.601841
Region XII	6.540952
Region XIII	6.527808

```
[49]: # Plotting
plt.figure(figsize=(8, 5))
plt.barh(reg_veg.index, reg_veg['price_php'], color= colors[3])
plt.xticks(size=8)
plt.title('Average Price of Vegetables and Fruits Category per Region')
plt.xlabel('Price in (PHP)')
plt.ylabel('Region')
plt.grid(True)

# Adjust the layout
plt.tight_layout()

# Display the plot
plt.show()
```



1.0.18 14. What are the average price of cereals and tubers in Region IV-A and Region IV-B over the years?

```
[50]: food_prices['region'].unique()
```

```
[50]: array(['National Capital region', 'Region III', 'Region IX', 'Region VI',
        'Region VII', 'Region XI', 'Region XII',
        'Autonomous region in Muslim Mindanao',
        'Cordillera Administrative region', 'Region I', 'Region II',
        'Region IV-A', 'Region IV-B', 'Region V', 'Region VIII',
        'Region X', 'Region XIII'], dtype=object)
```

```
[51]: # Select the date wherein category is equals to cereals and region is equals to
        ↪reg IV-A
four_a_cereals = food_prices[(food_prices['category'] == 'cereals and tubers') &
                             (food_prices['region'] == 'Region IV-A')]

# Select the date wherein category is equals to cereals and region is equals to
        ↪reg IV-B
four_b_cereals = food_prices[(food_prices['category'] == 'cereals and tubers') &
                             (food_prices['region'] == 'Region IV-B')]
```

```
[52]: four_a_cereals = four_a_cereals.groupby('year').mean(numeric_only = True)
four_a_cereals
```

```
[52]:
```

	price_php	usdprice	month
year			
2008	27.812500	0.627597	6.500000
2009	30.314412	0.638200	6.205882
2010	36.423333	0.809633	6.500000
2012	34.638387	0.821187	6.387097
2013	38.268462	0.929577	3.076923
2014	46.918000	1.064620	9.800000
2015	44.301250	0.973000	6.500000
2016	56.212500	1.185646	6.500000
2017	52.528750	1.042817	6.500000
2018	58.894167	1.117112	6.500000
2019	58.817000	1.127270	3.000000
2020	51.171308	1.043214	8.501558
2021	51.443467	1.046735	6.500000
2022	50.458514	0.927832	6.500000
2023	54.434161	0.984402	4.000000

```
[53]: four_b_cereals = four_b_cereals.groupby('year').mean(numeric_only = True)
four_b_cereals
```

```
[53]:
```

	price_php	usdprice	month
year			
2008	29.109444	0.655825	6.500000
2009	27.146667	0.577185	7.000000
2010	41.194167	0.915471	6.500000
2012	36.071290	0.855239	6.387097
2013	38.886923	0.944154	3.076923
2014	54.575000	1.238810	9.800000
2015	50.202500	1.103888	6.500000
2016	63.353750	1.339850	6.500000
2017	53.595417	1.064225	6.500000
2018	58.070455	1.103486	6.545455
2019	62.113000	1.190740	3.000000
2020	47.481134	0.969117	8.601375
2021	43.284740	0.880582	6.500000
2022	57.379432	1.055638	6.500000
2023	63.543182	1.149082	4.000000

```
[54]: # X value from 2009-2023
years = range(2008,2024)

# Create and Display the plot in line chart
fig, ax = plt.subplots(figsize=(12, 6))
ax.plot(four_a_cereals.index, four_a_cereals['price_php'], label= 'Region_
↳IV-A',marker='o',color = colors[0])
```



```

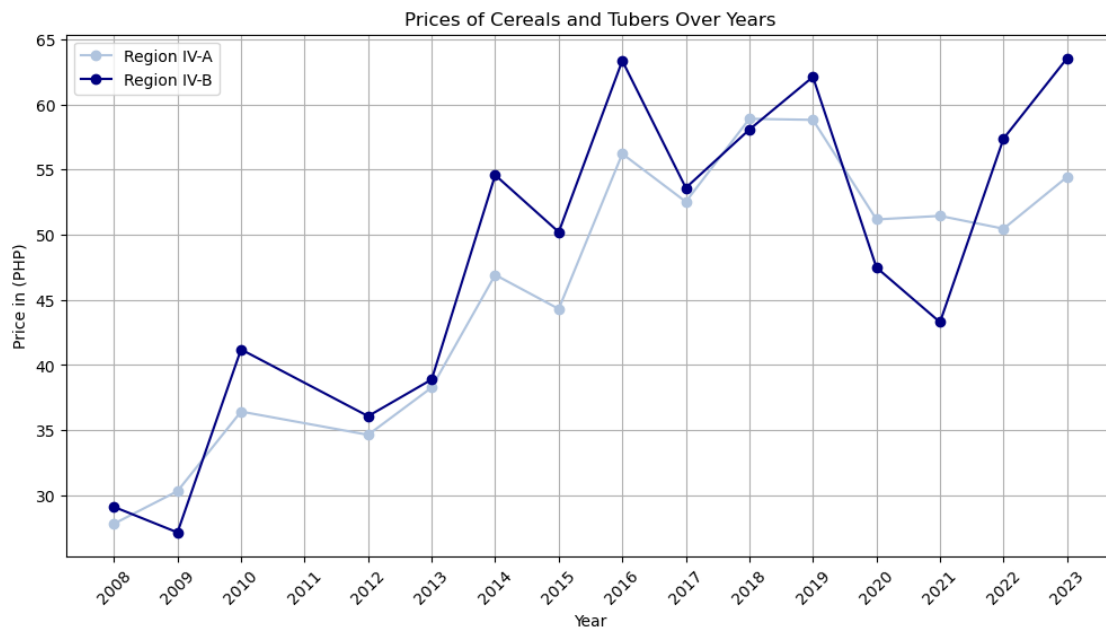
ax.plot(four_b_cereals.index, four_b_cereals['price_php'], label= 'Region_
↪IV-B',marker='o',color = colors[3])

# Set labels and title
ax.set_xlabel('Year')
ax.set_ylabel('Price in (PHP)')
ax.set_title('Prices of Cereals and Tubers Over Years')
plt.xticks(years, rotation=45)

# Add a legend
ax.legend()

# Display the plot
plt.grid(True)
plt.show()

```



1.0.19 15. What will be the price of Rice (regular, milled) in next five years?

```

[55]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression

X = rice_regular.index.values.reshape(-1, 1) # Years
y = rice_regular['price_php'].values # Prices

```

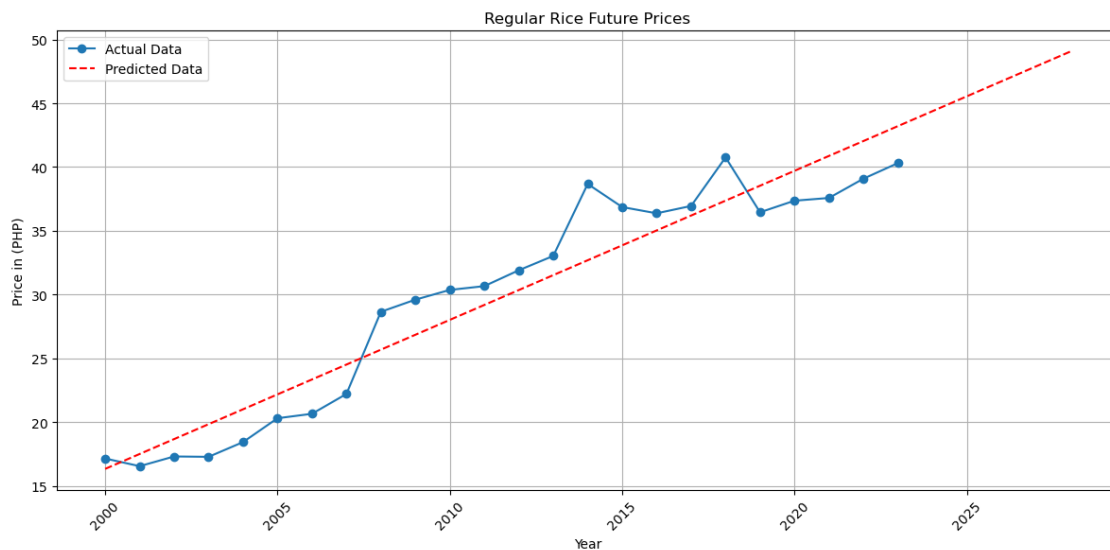
```

# Create and fit the linear regression model
model = LinearRegression()
model.fit(X, y)

# Predict prices for the actual years and the next 5 years
all_years = np.concatenate((X, np.arange(2024, 2029).reshape(-1, 1)))
predicted_prices = model.predict(all_years)

# Plot the actual data, linear regression line, and predicted data
plt.figure(figsize=(12, 6))
plt.plot(X, y, label='Actual Data', marker='o')
plt.plot(all_years, predicted_prices, label='Predicted Data',
         linestyle='dashed', color='red')
plt.xlabel('Year')
plt.ylabel('Price in (PHP)')
plt.title('Regular Rice Future Prices')
plt.xticks(rotation=45)
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.show()

```



[]: