NumPy based: Multidimensional Array Filtering Using Apartment

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1. Problem Statement:

- Define a 3D, 3x2x4 numpy array. We defined the indexes of this array as **floor**, apartment and room.
- Then, with filtering, we extracted the apartment numbers according to the magnets in the kitchens of the apartments in this building.

2. Initial Part

• We have defined a 3D, 3x2x4 numpy array named array_3D.

• Let's consider the first axis (dimension) of this array as "floors", the second axis as "apartments", and the third axis as "rooms".

[1. 2. 3. 4.]]

[[1. 2. 3. 4.] [2. 3. 4. 5.]]

- It has three floors ([0],[1], [2]) and two apartments(left[0], right[1]) on each floor. Each apartment has four rooms (living room[0], bedroom[1], kitchen[2], bathroom[3]).
- There are 3x2x4 = 24 rooms in total, and the values we give are the number of tables in each room.

```
In [ ]:
         print("Number of 'floors' on the first axis: ",array_3B.shape[0])
         print("Number of 'apartments' on the second axis: ",array 3B.shape[1])
         print("Number of 'rooms' on the third axis: ",array 3B.shape[2])
         Number of 'floors' on the first axis: 3
         Number of 'apartments' on the second axis: 2
         Number of 'rooms' on the third axis: 4
        Let's get, in one row, the number of tables (or fridge magnet) in the kitchens of each right apartment on each floor.
In [ ]:
         print("The number of magnets in the kitchen of the right apartments on each floor: ",array 3B[0:3,1,2])
         The number of magnets in the kitchen of the right apartments on each floor : [ 6 14 22]
        Let's list the apartments with more than 3 magnets in their kitchens.
In [ ]:
         array 3B filter = array 3B[array 3B[:,:,2] > 3]
         # Just to get the apartment numbers:
         array 3B apartments = np.delete(array 3B filter,[0,2,3],1)
         print("Apartments with more than 3 magnets in their kitchens", array 3B apartments.tolist())
         Apartments with more than 3 magnets in their kitchens [[5], [9], [13], [17], [21]]
        Bonus: Let's define the values of a 3-dimensional array as the sum of the indices of each respective cell. For example: array_3B will be [0,1,2] --> 3.
In [ ]:
         array 3B = np.empty([3,2,4])
         for first axis in np.arange(array 3B.shape[0]):
             for second axis in np.arange(array 3B.shape[1]):
                  for third axis in np.arange(array 3B.shape[2]):
                      array 3B [first axis, second axis, third axis] = first axis + second axis + third axis
         print("Our array :", array 3B)
         Our array : [[[0. 1. 2. 3.]
```

```
[[2. 3. 4. 5.]
[3. 4. 5. 6.]]]
```

1.Importing Libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

2.Importing and reading the Dataset

```
In [ ]: df=pd.read_csv('../Polynomial regression/data/Position_Salaries.csv')
    df
```

Out[]:		Position	Level	Salary
	0	Business Analyst	1	45000
	1	Junior Consultant	2	50000
	2	Senior Consultant	3	60000
	3	Manager	4	80000
	4	Country Manager	5	110000
	5	Region Manager	6	150000
	6	Partner	7	200000
	7	Senior Partner	8	300000
	8	C-level	9	500000
	9	CEO	10	1000000

3. Dividing the dataset into 2 components

Divide dataset into two components that is X and y.X will contain the Column between 1 and 2. y will contain the 2 columns.

```
In [ ]:
    X = df.iloc[:, 1:2].values
    y = df.iloc[:, 2].values
```

4. Fitting Linear Regression to the dataset

Fitting the linear Regression model On two components.

```
In []: # Fitting Linear Regression to the dataset
    from sklearn.linear_model import LinearRegression
    lin = LinearRegression()

Out[]: LinearRegression()
```

5. Fitting Polynomial Regression to the dataset

Fitting the Polynomial Regression model on two components X and y.

```
In [ ]: # Fitting Polynomial Regression to the dataset
from sklearn.preprocessing import PolynomialFeatures

poly = PolynomialFeatures(degree = 4)
X_poly = poly.fit_transform(X)

poly.fit(X_poly, y)
lin2 = LinearRegression()
lin2.fit(X_poly, y)
```

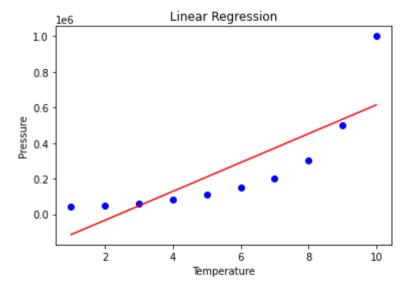
```
Out[]: LinearRegression()
```

6.In this step, we are Visualising the Linear Regression results using a scatter plot.

```
In [ ]:
# Visualising the Linear Regression results
plt.scatter(X, y, color = 'blue')

plt.plot(X, lin.predict(X), color = 'red')
plt.title('Linear Regression')
plt.xlabel('Temperature')
plt.ylabel('Pressure')

plt.show()
```

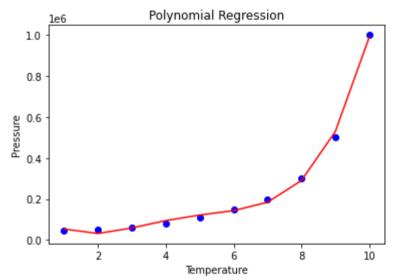


7. Visualising the Polynomial Regression results using a scatter plot.

```
In [ ]:  # Visualising the Polynomial Regression results
    plt.scatter(X, y, color = 'blue')
```

```
plt.plot(X, lin2.predict(poly.fit_transform(X)), color = 'red')
plt.title('Polynomial Regression')
plt.xlabel('Temperature')
plt.ylabel('Pressure')

plt.show()
```



8.Predicting new results with both Linear and Polynomial Regression. Note that the input variable must be in a numpy 2D array.

```
In []: # Predicting a new result with Linear Regression after converting predict variable to 2D array pred = 110.0 predarray = np.array([[pred]]) lin.predict(predarray)

Out[]: array([8701333.3333333])

In []: # Predicting a new result with Polynomial Regression after converting predict variable to 2D array pred2 = 110.0
```

```
pred2array = np.array([[pred2]])
lin2.predict(poly.fit_transform(pred2array))

Out[]: array([1.10869084e+11])
```