1. Write a program to distinguish between Array Indexing and Fancy Indexing.

```
import numpy as np

#array indexing
arr=np.array([2,4,6,8,10,12])
print(arr)

#fancy indexing
fancy_array = np.array([2,4,6,8,10,12])
selected_index = fancy_array[[0,2,3,5]]
print(selected_index)

[ 2  4  6  8 10 12]
[ 2  6  8 12]
```

2.Execute the 2D array Slicing.

```
import numpy as np
arr=np.array([[1,2,3,4,5,6],[7,8,9,10,11,12]])
print(arr[1][1:5])
[ 8  9 10 11]
```

3. Create the 5-Dimensional arrays using 'ndmin'.

```
import numpy as np
arr1=np.array([9,8,7,6],ndmin=5)
print("The 5 dimensional array using ndmin is : ",arr1)
The 5 dimensional array using ndmin is : [[[[[9 8 7 6]]]]]
```

4. Reshape the array from 1-D to 2-D array.

```
import numpy as np
arr=np.array([0,9,8,7,6,5,4,3,2,1])
print("The original array :",arr)
reshapearr=arr.reshape(2,5)
print("Array after reshapeing : \n",reshapearr)

The original array : [0 9 8 7 6 5 4 3 2 1]
Array after reshapeing :
  [[0 9 8 7 6]
  [5 4 3 2 1]]
```

5.Perform the Stack functions in Numpy arrays – Stack(), hstack(), vstack(), and dstack().

```
import numpy as np
arr1=np.array([00,11,22,33,44,55,66,77,88,99])
```

```
print("First array : ",arr1)
arr2=np.array([10,20,30,40,50,60,70,80,90,100])
print("Second array : ",arr2)
#Using stack
sarr=np.stack((arr1,arr2))
print("\nArrays after using stack : \n",sarr)
#Usina hstack
harr=np.hstack((arr1,arr2))
print("\nArrays after using hstack : \n",harr)
#Using vstack
varray=np.vstack((arr1,arr2))
print("\nArrays after using vstack : \n",varray)
#Using dstack
darray=np.dstack((arr1,arr2))
print("\nArrays after using dstack :\n",darray)
First array : [ 0 11 22 33 44 55 66 77 88 99]
Second array: [ 10 20 30 40 50 60 70 80 90 100]
Arrays after using stack:
[ 0 11 22 33 44 55 66 77 88 99]
 [ 10 20 30 40 50 60 70 80 90 100]]
Arrays after using hstack:
[ 0 11 22 33 44 55 66 77 88 99 10 20 30 40 50 60 70
80
 90 1001
Arrays after using vstack:
 [[ 0 11 22 33 44 55 66 77 88 99]
 [ 10 20 30 40 50 60 70 80 90 100]]
Arrays after using dstack:
 [[[0 10]
 [ 11 20]
  [ 22 30]
  [ 33 40]
  [ 44 50]
  [ 55 60]
  [ 66
      70]
  [ 77 80]
  [ 88 90]
  [ 99 100]]]
```

6.Perform the searchsort method in Numpy array.

```
import numpy as np
arr=np.array([12,23,34,45,56,67,78])

sarr=np.searchsorted(arr,[45,56,78,12])
print("Search Sort : ",sarr)

sarray=np.searchsorted(arr,23,side='left')
print("(using side) : ",sarray)

Search Sort : [3 4 6 0]
(using side) : 1
```

7.Create Numpy Structured array using your domain features.

8. Create Data frame using List and Dictionary.

```
import pandas as pd
#Dataframe using Dictionary
df={
    'Books':['Aarachar','Harry Potter 1','Percy Jackson','Chemmeen'],
    'Year': [2012, 1997, 2010, 1956]
mybooks=pd.DataFrame(df)
print(mybooks)
#DataFrame using list
mylist=[2012,1997,2010,1956]
mydata=pd.DataFrame(mylist,index=['a','b','c','d'],
                    columns=['Years'])
print("\n\n", mydata)
            Books Year
0
         Aarachar 2012
1 Harry Potter 1 1997
2
  Percy Jackson 2010
3
         Chemmeen 1956
    Years
    2012
a
b
    1997
```

```
c 2010
d 1956
```

9.Create Data frame on your Domain area and perform the following operations to find and eliminate the missing data from the dataset. • isnull() • notnull() • dropna() • fillna() • replace() • interpolate()

```
import pandas as pd
import numpy as np
data = {
    'Year': [2010, 2011, 2012, 2013, 2014],
    'Crop Yield': [500, 600, np.nan, 700, 750],
    'Rainfall mm': [800, 850, 900, np.nan, 950]
}
df = pd.DataFrame(data)
# Check for missing data in the entire DataFrame
print(df.isnull())
# Check for non-missing data in the entire DataFrame
print(df.notnull())
# Drop rows with any missing data
df.dropna(axis=0, inplace=True)
# Drop columns with any missing data
df.dropna(axis=1, inplace=True)
# Fill missing values in the 'Crop_Yield' column with the mean value
of that column
mean yield = df['Crop Yield'].mean()
df['Crop Yield'].fillna(mean yield, inplace=True)
# Fill missing values in the 'Rainfall mm' column with a specific
value, e.g., 0
df['Rainfall mm'].fillna(0, inplace=True)
# Replace a specific value in the DataFrame
df.replace(700, 720, inplace=True)
# Replace multiple values in the DataFrame
df.replace({600: 620, 750: 770}, inplace=True)
```

```
# Interpolate missing values in a column using linear interpolation
df['Rainfall mm'].interpolate(method='linear', inplace=True)
   Year Crop Yield Rainfall mm
   False
               False
                            False
1
  False
               False
                            False
  False
                            False
               True
3
  False
               False
                            True
4 False
               False
                            False
  Year Crop Yield Rainfall mm
0
  True
              True
                           True
1
  True
              True
                           True
  True
              False
                           True
3
  True
              True
                           False
4 True
              True
                            True
```

Q10. Perform the Hierarchical Indexing in the above created dataset.

```
import pandas as pd
import numpy as np
# Create the DataFrame
data = {
    'Year': [2010, 2011, 2012, 2013, 2014],
    'Crop Yield': [500, 600, np.nan, 700, 750],
    'Rainfall mm': [800, 850, 900, np.nan, 950]
}
df = pd.DataFrame(data)
# Perform Hierarchical Indexing
df.set_index(['Year', 'Crop_Yield'], inplace=True)
# Access data using hierarchical index
value = df.loc[(2010, 500)]
# Select all data for the year 2011
year 2011 data = df.xs(key=2011, level='Year')
# Select all data for a specific crop yield (e.g., 600)
crop yield 600 data = df.xs(key=600, level='Crop Yield')
print("Hierarchical Indexing Result:")
print(df)
print("\nAccessing Data:")
print("Value at (2010, 500):", value)
print("\nData for Year 2011:")
```

```
print(year_2011_data)
print("\nData for Crop Yield 600:")
print(crop_yield_600_data)
Hierarchical Indexing Result:
                  Rainfall_mm
Year Crop_Yield
2010 500.0
                        800.0
2011 600.0
                        850.0
2012 NaN
                        900.0
2013 700.0
                          NaN
2014 750.0
                        950.0
Accessing Data:
Value at (2010, 500): Rainfall_mm
                                       800.0
Name: (2010, 500.0), dtype: float64
Data for Year 2011:
            Rainfall mm
Crop_Yield
600.\overline{0}
                   850.0
Data for Crop Yield 600:
      Rainfall mm
Year
2011
            850.0
```