| in [1]: | impo | mport pandas as pd | | | | | | | | | | | | |
|---------|------|--|---------|----|------|----|------|----|------|--|--|--|--|--|
| [n [2]: | p=pd | e=pd.read_excel("MIDMARKS-MINOR1-EXAM.xlsx") | | | | | | | | | | | | |
| n [3]: | р | р | | | | | | | | | | | | |
| ut[3]: | | S.NO | SECTION | DV | M-II | PP | BEEE | FL | FIMS | | | | | |
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | | | | | |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | | | | | |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | | | | | |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | | | | | |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | | | | | |
| | ••• | | | | | | | | | | | | | |
| | 475 | 476 | NaN | 18 | 2 | 12 | 3 | 17 | 15 | | | | | |
| | 476 | 477 | NaN | 20 | 6 | 16 | 11 | 20 | 14 | | | | | |
| | 477 | 478 | NaN | 20 | NaN | 18 | 13 | 20 | 18 | | | | | |
| | 478 | 479 | NaN | 20 | 20 | 5 | 19 | 18 | 14 | | | | | |
| | 479 | 480 | NaN | 20 | 16 | 18 | 19 | 20 | 19 | | | | | |

480 rows × 8 columns

Importing pandas and reading Excel file data

In [4]: p.head(91)

| Out[4]: | | S.NO | SECTION | DV | M-II | PP | BEEE | FL | FIMS |
|---------|-----|------|---------|----|------|----|------|----|------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 |
| | ••• | | | | | | | | |
| | 86 | 87 | BETA | 17 | 18 | 19 | 20 | 20 | 18 |
| | 87 | 88 | BETA | 13 | 17 | 14 | 19 | 15 | 17 |
| | 88 | 89 | BETA | 2 | 17 | 0 | 3 | 15 | 2 |
| | 89 | 90 | BETA | 10 | 6 | 15 | 10 | 15 | 10 |
| | 90 | 91 | ВЕТА | 17 | 19 | 20 | 17 | 20 | 18 |

91 rows × 8 columns

In [5]:

| | | | _ |
|--------|------|---|-------|
| \cap | 1.11 | + | ١. |
| \cup | u | u | Ι. |

| | S.NO | SECTION | DV | M-II | PP | BEEE | FL | FIMS |
|-----|------|---------|----|------|----|------|----|------|
| 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 |
| 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 |
| 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 |
| 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 |
| 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 |
| ••• | | | | | | | | |
| 475 | 476 | NaN | 18 | 2 | 12 | 3 | 17 | 15 |
| 476 | 477 | NaN | 20 | 6 | 16 | 11 | 20 | 14 |
| 477 | 478 | NaN | 20 | NaN | 18 | 13 | 20 | 18 |
| 478 | 479 | NaN | 20 | 20 | 5 | 19 | 18 | 14 |
| 479 | 480 | NaN | 20 | 16 | 18 | 19 | 20 | 19 |

480 rows × 8 columns

In [6]: p.info()

```
<class 'pandas.core.frame.DataFrame'>
      RangeIndex: 480 entries, 0 to 479
      Data columns (total 8 columns):
       # Column Non-Null Count Dtype
                 -----
          S.NO 480 non-null
       0
                                 int64
       1
         SECTION 439 non-null object
       2 DV 479 non-null object
       3 M-II 477 non-null object
       4
                  480 non-null object
         BEEE 478 non-null object
       5
         FL
                  479 non-null object
       6
       7
          FIMS
                 480 non-null
                                 object
      dtypes: int64(1), object(7)
      memory usage: 30.1+ KB
In [7]:
       p['DV'] = pd.to_numeric(p['DV'], errors='coerce').astype('Int64')
       p['M-II'] = pd.to_numeric(p['M-II'], errors='coerce').astype('Int64')
       p['PP'] = pd.to_numeric(p['PP'], errors='coerce').astype('Int64')
       p['BEEE'] = pd.to_numeric(p['BEEE'], errors='coerce').astype('Int64')
       p['FL'] = pd.to_numeric(p['FL'], errors='coerce').astype('Int64')
       p['FIMS'] = pd.to_numeric(p['FIMS'], errors='coerce').astype('Int64')
In [8]:
       p.info()
       p["Total"]=p["DV"]+p["M-II"]+p["PP"]+p["BEEE"]+p["FL"]+p["FIMS"]
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 480 entries, 0 to 479
      Data columns (total 8 columns):
       # Column Non-Null Count Dtype
      --- ----- -----
       0 S.NO
                  480 non-null
                                 int64
       1 SECTION 439 non-null object
       2 DV 472 non-null Int64
                 465 non-null Int64
       3 M-II
          PP
                 470 non-null Int64
       5
          BEEE
                 464 non-null Int64
                  470 non-null Int64
       6 FL
          FIMS
                  466 non-null
                                 Int64
      dtypes: Int64(6), int64(1), object(1)
      memory usage: 32.9+ KB
```

| Out[8]: | | S.NO | SECTION | DV | M-II | PP | BEEE | FL | FIMS | Total |
|---------|-----|------|---------|----|-----------|----|------|----|------|-----------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 |
| | ••• | | ••• | | ••• | | | | | ••• |
| | 475 | 476 | NaN | 18 | 2 | 12 | 3 | 17 | 15 | 67 |
| | 476 | 477 | NaN | 20 | 6 | 16 | 11 | 20 | 14 | 87 |
| | 477 | 478 | NaN | 20 | <na></na> | 18 | 13 | 20 | 18 | <na></na> |
| | 478 | 479 | NaN | 20 | 20 | 5 | 19 | 18 | 14 | 96 |
| | 479 | 480 | NaN | 20 | 16 | 18 | 19 | 20 | 19 | 112 |

480 rows × 9 columns

```
In [9]: p.rename(columns={'Total':'TOTAL'},inplace=True)
p.rename(columns={'M-II':'M2'},inplace=True)
p
```

| Out[9]: | | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL |
|---------|-----|------|---------|----|-----------|----|------|----|------|-----------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 |
| | ••• | | | | | | | | | |
| | 475 | 476 | NaN | 18 | 2 | 12 | 3 | 17 | 15 | 67 |
| | 476 | 477 | NaN | 20 | 6 | 16 | 11 | 20 | 14 | 87 |
| | 477 | 478 | NaN | 20 | <na></na> | 18 | 13 | 20 | 18 | <na></na> |
| | 478 | 479 | NaN | 20 | 20 | 5 | 19 | 18 | 14 | 96 |
| | 479 | 480 | NaN | 20 | 16 | 18 | 19 | 20 | 19 | 112 |

480 rows × 9 columns

Renaming column 'M-II' to 'M2' in dataframe

In [10]: p.fillna(0)

| _ | | Га | | |
|-------|----|-----|----------|--|
| () i | ut | 1.7 | α | |
| 01 | ич | 1 - | v i | |

| | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL |
|-----|------|---------|----|----|----|------|----|------|-------|
| 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 |
| 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 |
| 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 |
| 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 |
| 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 |
| ••• | | | | | | | | | |
| 475 | 476 | 0 | 18 | 2 | 12 | 3 | 17 | 15 | 67 |
| 476 | 477 | 0 | 20 | 6 | 16 | 11 | 20 | 14 | 87 |
| 477 | 478 | 0 | 20 | 0 | 18 | 13 | 20 | 18 | 0 |
| 478 | 479 | 0 | 20 | 20 | 5 | 19 | 18 | 14 | 96 |
| 479 | 480 | 0 | 20 | 16 | 18 | 19 | 20 | 19 | 112 |

480 rows × 9 columns

In [11]: p.head(10)

Out[11]:

| | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL |
|---|------|---------|----|----|-----------|------|----|------|-----------|
| 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 |
| 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 |
| 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 |
| 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 |
| 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 |
| 5 | 6 | ALPHA | 17 | 16 | 18 | 10 | 15 | 9 | 85 |
| 6 | 7 | ALPHA | 15 | 10 | 20 | 20 | 15 | 14 | 94 |
| 7 | 8 | ALPHA | 17 | 17 | 19 | 20 | 19 | 13 | 105 |
| 8 | 9 | ALPHA | 10 | 18 | <na></na> | 20 | 19 | 15 | <na></na> |
| 9 | 10 | ALPHA | 18 | 19 | 20 | 20 | 20 | 15 | 112 |

In [12]: p = p.fillna(-1)

In [13]: p.head(10)

```
Out[13]:
            S.NO SECTION DV M2 PP BEEE FL FIMS TOTAL
         0
               1
                    ALPHA
                            12
                                 0
                                    17
                                           9 19
                                                    15
                                                           72
         1
               2
                    ALPHA
                            19
                                12
                                    16
                                          16 18
                                                     3
                                                           84
         2
               3
                    ALPHA
                            18
                                14
                                    18
                                          18 18
                                                    16
                                                          102
         3
               4
                    ALPHA
                            15
                                 9
                                    19
                                          17 19
                                                    15
                                                           94
         4
               5
                    ALPHA
                            18
                                17
                                    19
                                          19 20
                                                    18
                                                          111
         5
               6
                    ALPHA
                            17
                                16
                                    18
                                          10 15
                                                           85
         6
               7
                    ALPHA
                            15
                                10
                                    20
                                          20 15
                                                    14
                                                           94
         7
               8
                    ALPHA
                            17
                                17
                                    19
                                          20 19
                                                    13
                                                          105
         8
               9
                    ALPHA
                            10
                                18
                                    -1
                                          20 19
                                                    15
                                                           -1
              10
                    ALPHA
                            18
                                19
                                    20
                                          20 20
                                                    15
                                                          112
In [16]: p.loc[600:630]
Out[16]:
           S.NO SECTION DV M2 PP BEEE FL FIMS TOTAL
         p.loc[560:570]
In [17]:
Out[17]:
           S.NO SECTION DV M2 PP BEEE FL FIMS TOTAL
In [18]: p.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 480 entries, 0 to 479
        Data columns (total 9 columns):
                     Non-Null Count Dtype
        #
            Column
         0
            S.NO
                     480 non-null
                                     int64
            SECTION 480 non-null
         1
                                     object
         2
            DV
                     480 non-null
                                     Int64
         3
            M2
                     480 non-null
                                     Int64
         4
            PP
                     480 non-null
                                     Int64
            BEEE
                     480 non-null
                                     Int64
         5
         6
            FL
                     480 non-null
                                     Int64
         7
             FIMS
                     480 non-null
                                     Int64
             TOTAL
                     480 non-null
                                     Int64
        dtypes: Int64(7), int64(1), object(1)
        memory usage: 37.2+ KB
```

In [19]: p

| Out[19]: | | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL |
|----------|-----|------|---------|----|----|----|------|----|------|-------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 |
| | ••• | | | | | | | | | |
| | 475 | 476 | -1 | 18 | 2 | 12 | 3 | 17 | 15 | 67 |
| | 476 | 477 | -1 | 20 | 6 | 16 | 11 | 20 | 14 | 87 |
| | 477 | 478 | -1 | 20 | -1 | 18 | 13 | 20 | 18 | -1 |
| | 478 | 479 | -1 | 20 | 20 | 5 | 19 | 18 | 14 | 96 |
| | 479 | 480 | -1 | 20 | 16 | 18 | 19 | 20 | 19 | 112 |

480 rows × 9 columns

| In [20]: | p["percentage"]=(p["TOTAL"]/120)*100 |
|----------|--------------------------------------|
| In [21]: | p |

| Out[21]: | | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL | percentage |
|----------|-----|------|---------|----|----|----|------|----|------|-------|------------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 | 60.0 |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 | 70.0 |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 | 85.0 |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 | 78.333333 |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 | 92.5 |
| | ••• | | | | | | | | | | |
| | 475 | 476 | -1 | 18 | 2 | 12 | 3 | 17 | 15 | 67 | 55.833333 |
| | 476 | 477 | -1 | 20 | 6 | 16 | 11 | 20 | 14 | 87 | 72.5 |
| | 477 | 478 | -1 | 20 | -1 | 18 | 13 | 20 | 18 | -1 | -0.833333 |
| | 478 | 479 | -1 | 20 | 20 | 5 | 19 | 18 | 14 | 96 | 80.0 |
| | 479 | 480 | -1 | 20 | 16 | 18 | 19 | 20 | 19 | 112 | 93.333333 |

480 rows × 10 columns

Calculating percentage based on 'Total' column values

```
In [22]: p["grade"]=((p["TOTAL"]/120)*10).round()
In [23]: p
```

BEEE FL FIMS TOTAL percentage grade Out[23]: S.NO SECTION DV **M2** PP **ALPHA** 60.0 6.0 **ALPHA** 70.0 7.0 85.0 **ALPHA** 8.0 **ALPHA** 78.333333 0.8 92.5 **ALPHA** 9.0 -1 3 17 55.833333 6.0 -1 72.5 7.0 -1 -1 -0.833333 -0.0 -1 -1 0.08 0.8

19 20

93.333333

9.0

480 rows × 11 columns

-1

16 18

```
In [24]:

def assign_grade(percentage):
    if percentage >= 90:
        return 'A'
    elif percentage >= 80:
        return 'B'
    elif percentage >= 70:
        return 'C'
    elif percentage >= 60:
        return 'D'
    else:
        return 'F'

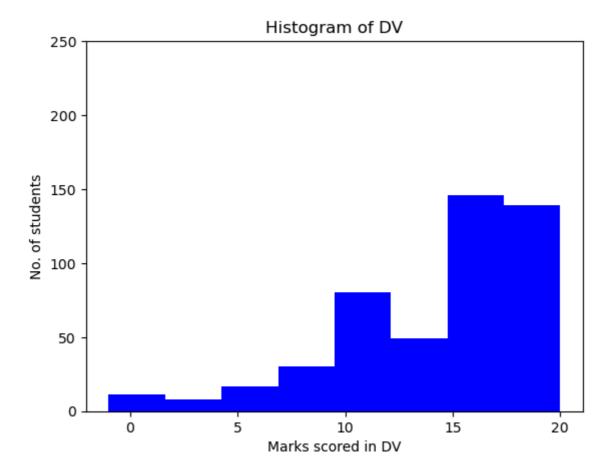
p['Grade'] = p['percentage'].apply(assign_grade)
p
```

| Out[24]: | | S.NO | SECTION | DV | M2 | PP | BEEE | FL | FIMS | TOTAL | percentage | grade | Grade |
|----------|-----|------|---------|----|----|----|------|----|------|-------|------------|-------|-------|
| | 0 | 1 | ALPHA | 12 | 0 | 17 | 9 | 19 | 15 | 72 | 60.0 | 6.0 | D |
| | 1 | 2 | ALPHA | 19 | 12 | 16 | 16 | 18 | 3 | 84 | 70.0 | 7.0 | C |
| | 2 | 3 | ALPHA | 18 | 14 | 18 | 18 | 18 | 16 | 102 | 85.0 | 8.0 | В |
| | 3 | 4 | ALPHA | 15 | 9 | 19 | 17 | 19 | 15 | 94 | 78.333333 | 8.0 | C |
| | 4 | 5 | ALPHA | 18 | 17 | 19 | 19 | 20 | 18 | 111 | 92.5 | 9.0 | А |
| | ••• | | | | | | | | | | | | |
| | 475 | 476 | -1 | 18 | 2 | 12 | 3 | 17 | 15 | 67 | 55.833333 | 6.0 | F |
| | 476 | 477 | -1 | 20 | 6 | 16 | 11 | 20 | 14 | 87 | 72.5 | 7.0 | C |
| | 477 | 478 | -1 | 20 | -1 | 18 | 13 | 20 | 18 | -1 | -0.833333 | -0.0 | F |
| | 478 | 479 | -1 | 20 | 20 | 5 | 19 | 18 | 14 | 96 | 80.0 | 8.0 | В |
| | 479 | 480 | -1 | 20 | 16 | 18 | 19 | 20 | 19 | 112 | 93.333333 | 9.0 | А |

480 rows × 12 columns

Assigning grades based on percentage values in dataframe

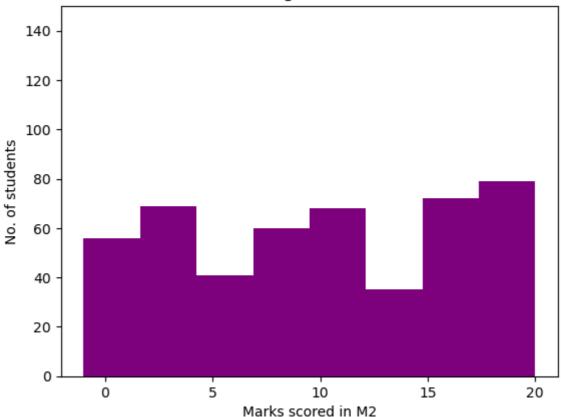
```
In [25]: import matplotlib.pyplot as plt
   plt.hist(p['DV'], color='blue', bins=8)
   plt.ylim(0, 250)
   plt.xlabel("Marks scored in DV")
   plt.ylabel("No. of students")
   plt.title("Histogram of DV")
   plt.show()
   #print(p['DV'].value_counts())
```



Creating histogram to visualize 'DV' subject marks distribution

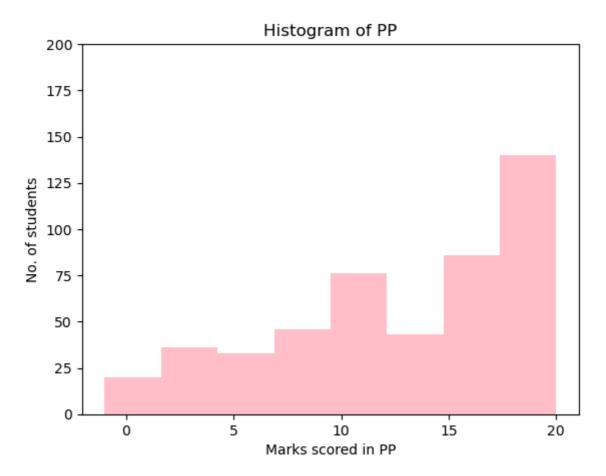
```
In [26]: plt.hist(p['M2'], color='purple', bins=8)
    plt.ylim(0, 150)
    plt.xlabel("Marks scored in M2")
    plt.ylabel("No. of students")
    plt.title("Histogram of M2")
    plt.show()
    #print(p['M2'].value_counts())
```





Creating histogram to visualize 'M2' subject marks distribution

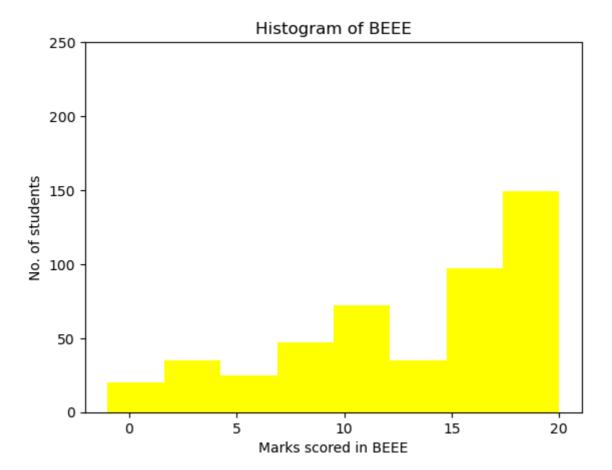
```
In [27]: plt.hist(p['PP'], color='pink', bins=8)
    plt.ylim(0, 200)
    plt.xlabel("Marks scored in PP")
    plt.ylabel("No. of students")
    plt.title("Histogram of PP")
    plt.show()
```



In []:

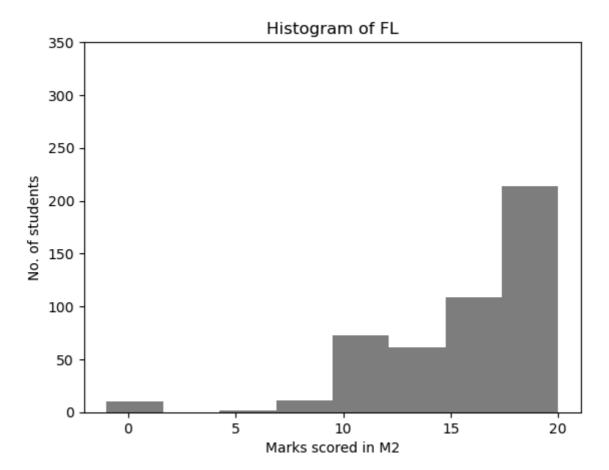
Creating histogram to visualize 'PP' subject marks distribution

```
In [28]: plt.hist(p['BEEE'], color='yellow', bins=8)
    plt.ylim(0, 250)
    plt.xlabel("Marks scored in BEEE")
    plt.ylabel("No. of students")
    plt.title("Histogram of BEEE")
    plt.show()
```



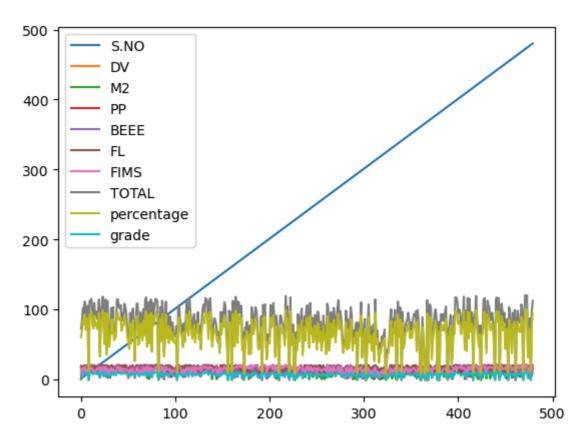
Creating histogram to visualize 'BEEE' subject marks distribution

```
In [29]: plt.hist(p['FL'], color='gray', bins=8)
    plt.ylim(0, 350)
    plt.xlabel("Marks scored in M2")
    plt.ylabel("No. of students")
    plt.title("Histogram of FL")
    plt.show()
```



Creating histogram to visualize 'FL' subject marks distribution

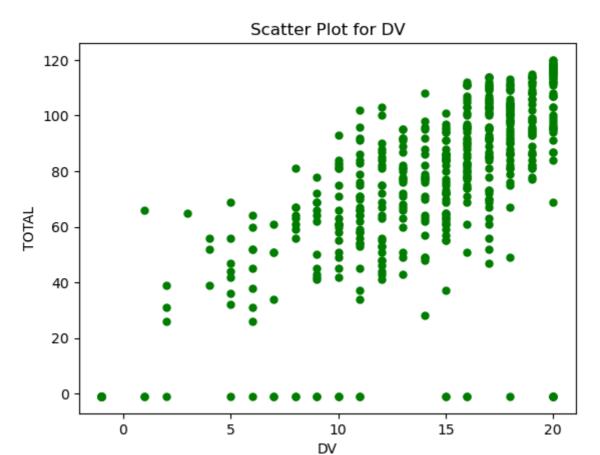
In [30]: p.plot()
 plt.show()



Plotting all columns in the dataframe for visualization

```
In [31]: p.plot.scatter(x = 'DV', y = 'TOTAL',color='green',s=25)
plt.title("Scatter Plot for DV")
```

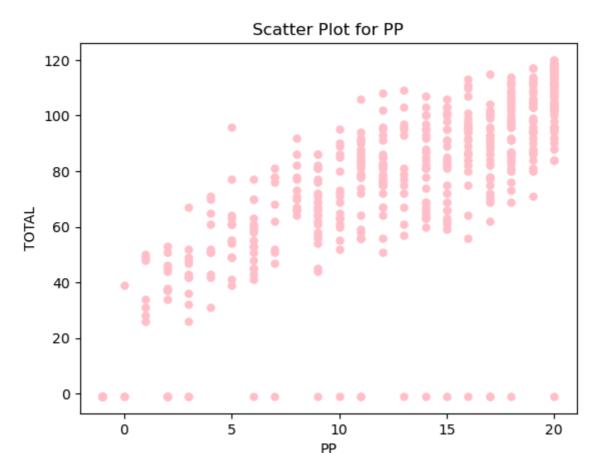
Out[31]: Text(0.5, 1.0, 'Scatter Plot for DV')



Creating scatter plot for 'DV' vs 'Total' values

```
In [32]: p.plot.scatter(x = 'PP', y = 'TOTAL',color='pink',s=25)
    plt.title("Scatter Plot for PP")
```

Out[32]: Text(0.5, 1.0, 'Scatter Plot for PP')

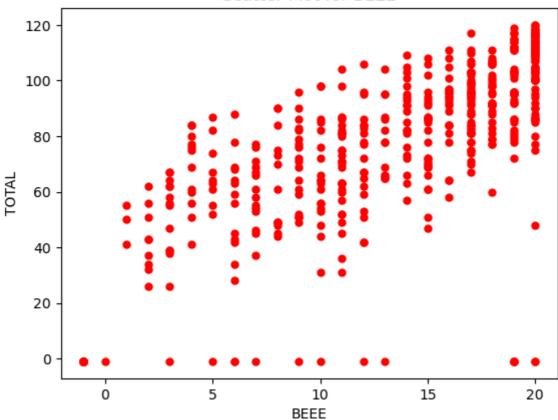


Creating scatter plot for 'PP' vs 'Total' values

```
In [33]: p.plot.scatter(x = 'BEEE', y = 'TOTAL',color='red',s=25)
    plt.title("Scatter Plot for BEEE")
```

Out[33]: Text(0.5, 1.0, 'Scatter Plot for BEEE')

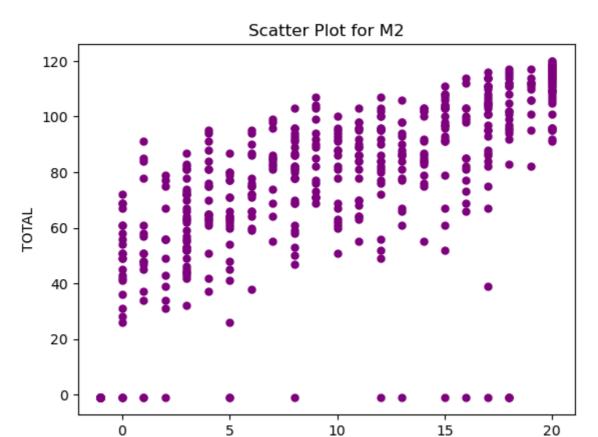
Scatter Plot for BEEE



Creating scatter plot for 'BEEE' vs 'Total' values

```
In [34]: p.plot.scatter(x = 'M2', y = 'TOTAL',color='purple',s=25)
plt.title("Scatter Plot for M2")
```

Out[34]: Text(0.5, 1.0, 'Scatter Plot for M2')

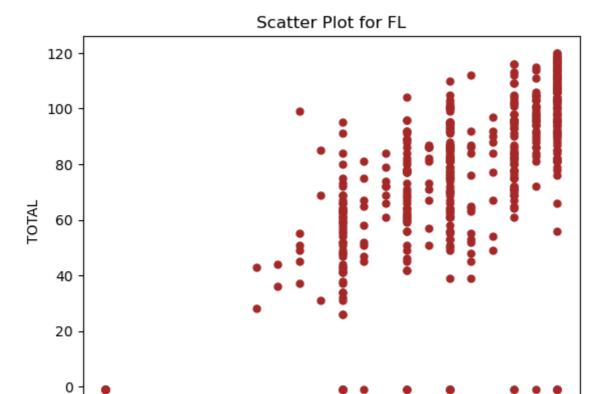


Creating scatter plot for 'M2' vs 'Total' values

Μ2

```
In [35]: p.plot.scatter(x = 'FL', y = 'TOTAL',color='brown',s=25)
plt.title("Scatter Plot for FL")
```

Out[35]: Text(0.5, 1.0, 'Scatter Plot for FL')



10

FL

15

20

Creating scatter plot for 'FL' vs 'Total' values

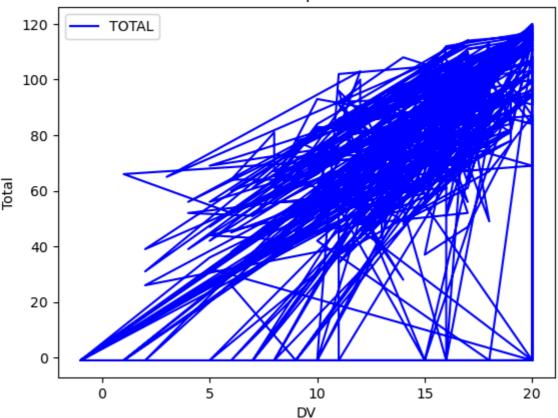
5

```
In [36]: p.plot.line(x='DV',y='TOTAL',color='blue')
plt.title("Line Graph of DV")
plt.ylabel("Total")
```

Out[36]: Text(0, 0.5, 'Total')

0

Line Graph of DV

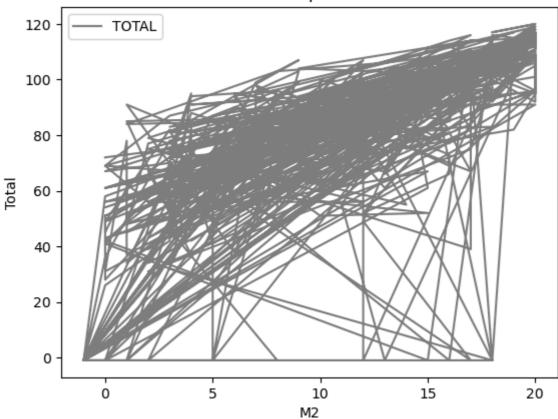


Plotting line graph of 'DV' vs 'Total' values

```
In [37]: p.plot.line(x='M2',y='TOTAL',color='gray')
   plt.title("Line Graph of M2")
   plt.ylabel("Total")
```

Out[37]: Text(0, 0.5, 'Total')

Line Graph of M2

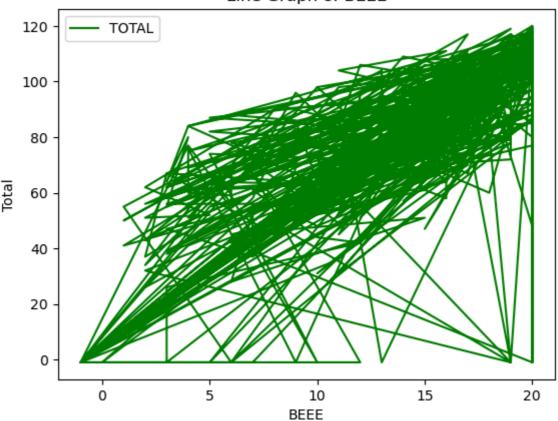


Plotting line graph of 'M2' vs 'Total' values

```
In [38]: p.plot.line(x='BEEE',y='TOTAL',color='green')
   plt.title("Line Graph of BEEE")
   plt.ylabel("Total")
```

Out[38]: Text(0, 0.5, 'Total')

Line Graph of BEEE

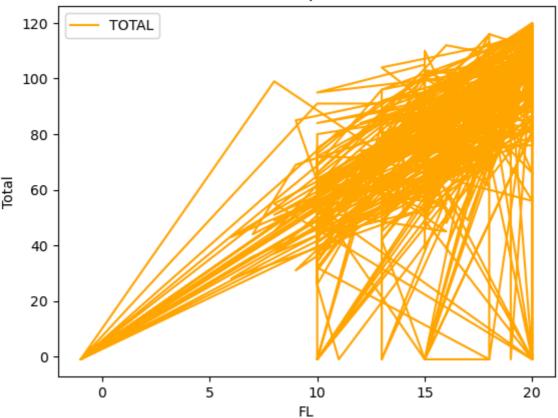


Plotting line graph of 'BEEE' vs 'Total' values

```
In [39]: p.plot.line(x='FL',y='TOTAL',color='orange')
plt.title("Line Graph of FL")
plt.ylabel("Total")
```

Out[39]: Text(0, 0.5, 'Total')

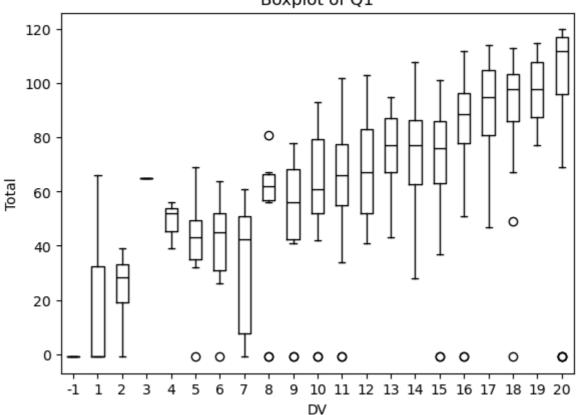




Plotting line graph of 'FL' vs 'Total' values

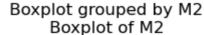
```
In [40]: p.boxplot(by='DV', column =['TOTAL'], grid = False,color='black')
plt.title("Boxplot of Q1")
plt.ylabel("Total")
plt.show()
```

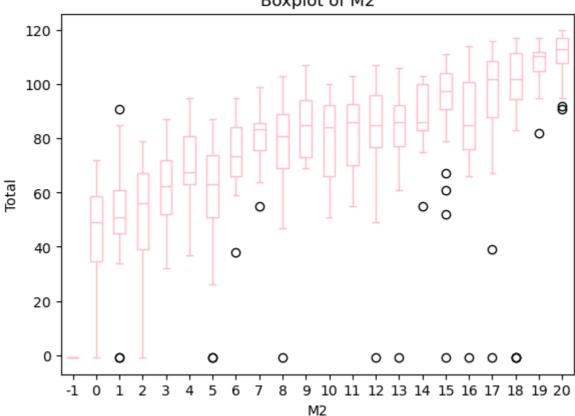
Boxplot grouped by DV Boxplot of Q1



Creating boxplot to visualize 'Total' distribution by 'DV'

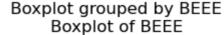
```
In [41]: p.boxplot(by='M2', column =['TOTAL'], grid = False,color='pink')
plt.title("Boxplot of M2")
plt.ylabel("Total")
plt.show()
```

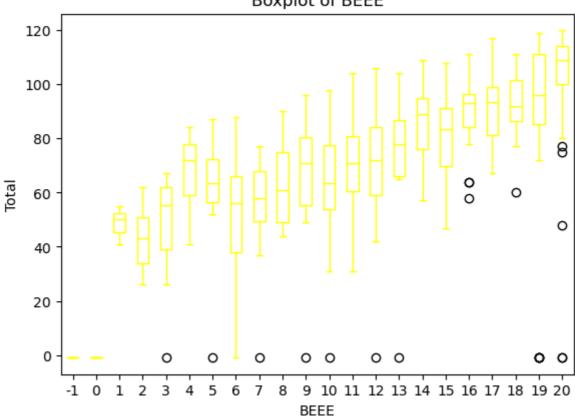




Creating boxplot to visualize 'Total' distribution by 'M2'

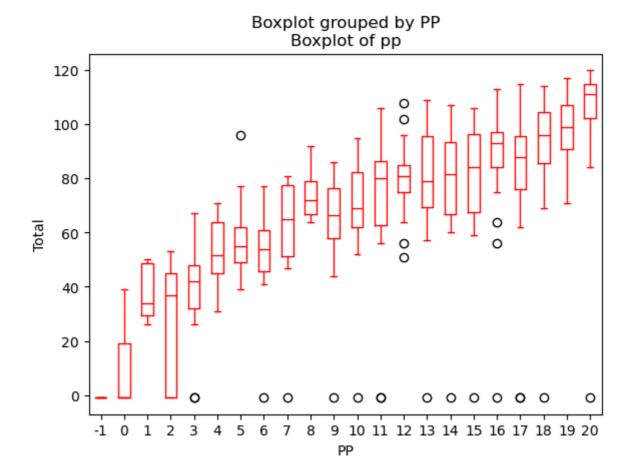
```
In [42]: p.boxplot(by='BEEE', column =['TOTAL'], grid = False,color='yellow')
    plt.title("Boxplot of BEEE")
    plt.ylabel("Total")
    plt.show()
```





Creating boxplot to visualize 'Total' distribution by 'BEEE'

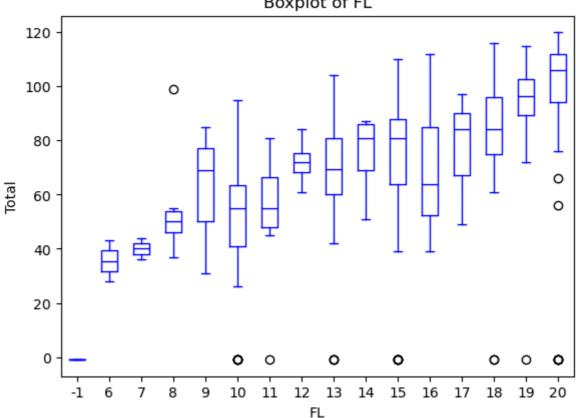
```
In [43]: p.boxplot(by='PP', column =['TOTAL'], grid = False,color='red')
plt.title("Boxplot of pp")
plt.ylabel("Total")
plt.show()
```



Creating boxplot to visualize 'Total' distribution by 'PP'

```
In [44]: p.boxplot(by='FL', column =['TOTAL'], grid = False,color='blue')
    plt.title("Boxplot of FL")
    plt.ylabel("Total")
    plt.show()
```

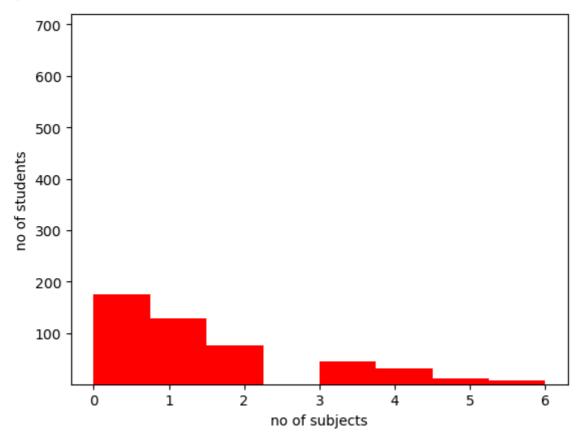




Creating boxplot to visualize 'Total' distribution by 'Fl'

```
In [45]:
          def failed(p):
               return ((p[['DV', 'M2', 'PP', 'BEEE', 'FL', 'FIMS']] < 10).sum(axis=1))</pre>
          p['backlog'] = failed(p)
          h = p.groupby('SECTION')['backlog'].sum()
          print(h)
        SECTION
        -1
                     61
                     43
        ALPHA
                     69
        BETA
                     73
        DELTA
        EPSILON
                     85
        GAMMA
                     93
        OMEGA
                    103
        SIGMA
                     97
                     39
        ZETA
        Name: backlog, dtype: Int64
 In [ ]:
In [46]:
          plt.hist(p['backlog'], color='red', bins=8)
          plt.ylim(1, 720)
```

```
plt.ylabel("no of students")
plt.xlabel("no of subjects")
plt.title("")
plt.show()
```

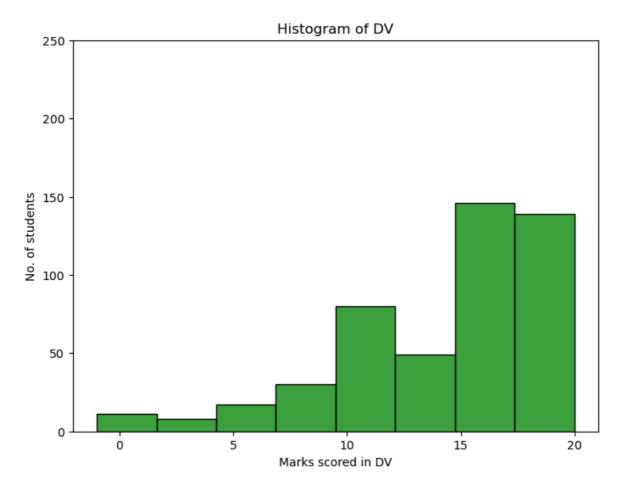


using seaborn no of subjects who are failed

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(8, 6))
sns.histplot(p['DV'], color='green', bins=8, kde=False)

plt.ylim(0, 250)
plt.xlabel("Marks scored in DV")
plt.ylabel("No. of students")
plt.title("Histogram of DV")
plt.show()
```



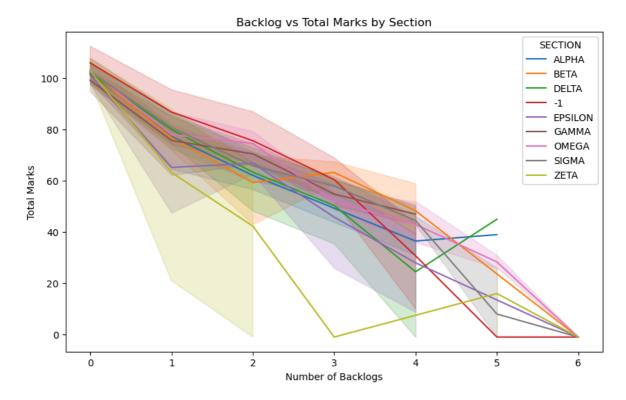
Creating histogram to visualize 'Dv' sbject marks distribution

```
In [48]: plt.figure(figsize=(10, 6))

sns.lineplot(x='backlog', y='TOTAL', hue='SECTION', data=p, markers=True)

plt.title('Backlog vs Total Marks by Section')
plt.xlabel('Number of Backlogs')
plt.ylabel('Total Marks')

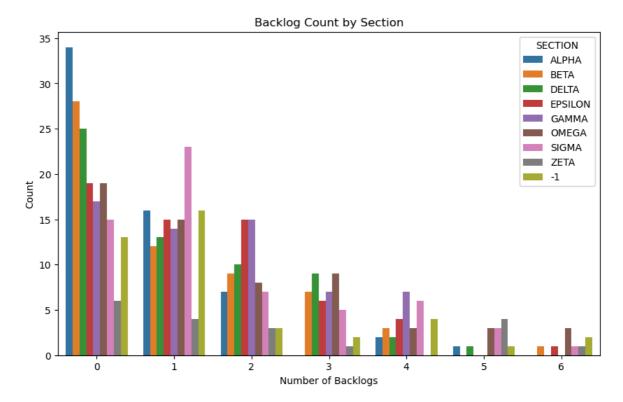
plt.show()
```



by using seaborn we are showcasing Backlog vs Total Marks by Section

In []:

by using seaborn we are showcasing Backlog vs Percentage by section

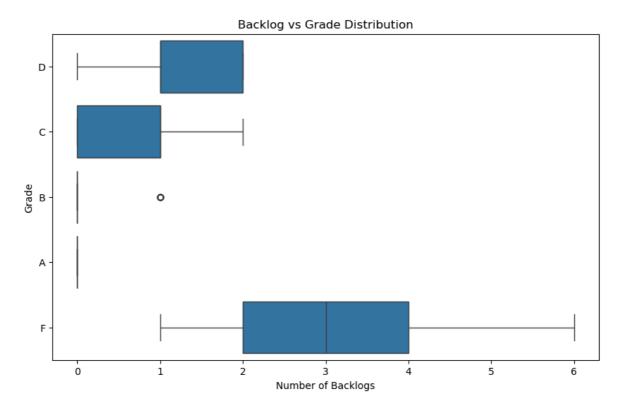


by using seaborn we are showcasing Backlog count by Section

```
In [50]: plt.figure(figsize=(10, 6))
    sns.boxplot(x='backlog', y='Grade', data=p)

plt.title('Backlog vs Grade Distribution')
    plt.xlabel('Number of Backlogs')
    plt.ylabel('Grade')

plt.show()
```



by using seaborn we are showcasing backlog vs grade distribution

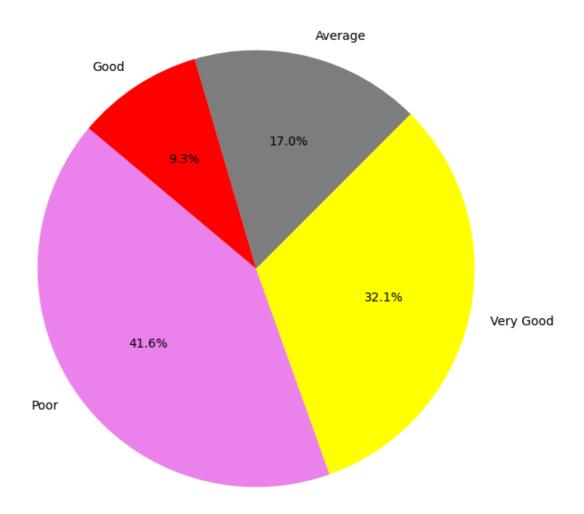
```
In [51]: import pandas as pd
          data = pd.read_excel("MIDMARKS.XLSX")
          subjects = ["PP"]
          data.dropna(inplace=True)
          for subject in subjects:
              data[subject] = pd.to_numeric(data[subject], errors='coerce').fillna(0)
          def assign_grade(percentage):
              if 18 <= percentage <= 20:</pre>
                  return 'Very Good'
              elif 13 <= percentage <= 14:</pre>
                  return 'Average'
              elif 13 <= percentage <= 17:</pre>
                  return 'Good'
              else:
                  return 'poor'
          data['PP_Grade'] = data['PP'].apply(assign_grade)
          print(data)
```

```
S.NO SECTION DV M-II
                                        PP BEEE
                                                  FL FIMS
                                                            PP Grade
                                                  19
        0
               1.0
                     ALPHA
                             12
                                   0
                                      17.0
                                               9
                                                       15
                                                                Good
        1
               2.0
                     ALPHA 19
                                  12 16.0
                                             16
                                                  18
                                                        3
                                                                 Good
        2
               3.0
                     ALPHA 18
                                  14 18.0
                                             18
                                                  18
                                                       16
                                                           Very Good
        3
               4.0
                     ALPHA 15
                                  9 19.0
                                             17
                                                  19
                                                       15
                                                           Very Good
        4
               5.0
                     ALPHA
                             18
                                  17 19.0
                                              19
                                                  20
                                                       18
                                                           Very Good
                             . .
                       . . .
                                 . . .
                                                  . .
        596
             597.0
                     SIGMA
                             20
                                  20 20.0
                                                  20
                                                       20
                                                           Very Good
                                             20
        597
             598.0
                     SIGMA
                             20
                                  20 20.0
                                                           Very Good
                                             19
                                                  19
                                                       18
        598
             599.0
                     SIGMA
                             20
                                  20 17.0
                                             17
                                                  19
                                                       18
                                                                 Good
        599
             600.0
                     SIGMA
                             14
                                  12 11.0
                                            9
                                                  18
                                                       17
                                                                 poor
            601.0
                                  19 20.0
        600
                     SIGMA 20
                                             18 18
                                                       19
                                                           Very Good
        [599 rows x 9 columns]
In [52]:
         import pandas as pd
         data = pd.read_excel("MIDMARKS.XLSX")
         subjects = ["DV"]
         data.dropna(inplace=True)
         for subject in subjects:
              data[subject] = pd.to_numeric(data[subject], errors='coerce').fillna(0)
         def assign_grade(percentage):
              if 18 <= percentage <= 20:</pre>
                  return 'Very Good'
              elif 13 <= percentage <= 14:</pre>
                  return 'Average'
              elif 13 <= percentage <= 17:</pre>
                  return 'Good'
              else:
                  return 'poor'
         data['DV_Grade'] = data['DV'].apply(assign_grade)
         print(data)
              S.NO SECTION
                               DV M-II
                                        PP BEEE
                                                  FL FIMS
                                                            DV Grade
        0
               1.0
                     ALPHA 12.0
                                     0
                                        17
                                               9
                                                  19
                                                       15
                                                                 poor
        1
               2.0
                     ALPHA 19.0
                                    12
                                        16
                                              16
                                                  18
                                                        3
                                                           Very Good
        2
               3.0
                     ALPHA 18.0
                                    14
                                        18
                                              18
                                                  18
                                                       16
                                                           Very Good
        3
                                     9
               4.0
                     ALPHA 15.0
                                        19
                                              17
                                                  19
                                                       15
                                                                 Good
        4
               5.0
                     ALPHA 18.0
                                        19
                                              19
                                    17
                                                  20
                                                       18
                                                           Very Good
               . . .
                       . . .
                              . . .
                                                  . .
        . .
                                   . . .
                                         . .
                                             . . .
                                                      . . .
                                                                  . . .
            597.0
        596
                     SIGMA 20.0
                                    20
                                        20
                                             20
                                                  20
                                                       20
                                                           Very Good
        597
             598.0
                     SIGMA
                             20.0
                                        20
                                             19
                                                  19
                                    20
                                                       18
                                                           Very Good
        598
             599.0
                      SIGMA
                             20.0
                                    20
                                        17
                                              17
                                                  19
                                                       18
                                                           Very Good
        599
             600.0
                             14.0
                                              9
                     SIGMA
                                    12
                                        11
                                                  18
                                                       17
                                                             Average
        600
             601.0
                     SIGMA
                             20.0
                                    19
                                        20
                                              18
                                                  18
                                                       19
                                                           Very Good
        [599 rows x 9 columns]
         data.head(2)
In [53]:
Out[53]:
             S.NO SECTION
                              DV
                                  M-II
                                        PP
                                            BEEE
                                                  FL FIMS
                                                            DV_Grade
          0
               1.0
                     ALPHA
                                                  19
                                                         15
                             12.0
                                     0
                                        17
                                                                  poor
               2.0
                     ALPHA 19.0
          1
                                    12
                                       16
                                               16 18
                                                          3 Very Good
```

```
In [54]:
        import pandas as pd
        import matplotlib.pyplot as plt
        df = pd.read_excel("MIDMARKS.XLSX")
        subjects = ["PP"]
        df.dropna(inplace=True) # Drop rows with missing values
        for subject in subjects:
            df[subject] = pd.to_numeric(df[subject], errors='coerce').fillna(0)
        def assign_grade(percentage):
            if 18 <= percentage <= 20:</pre>
                return 'Very Good'
            elif 15 <= percentage <= 17:</pre>
                return 'Average'
            elif 13 <= percentage <= 14:</pre>
                return 'Good'
            else:
                return 'Poor'
        df['Programming_skills'] = df['PP'].apply(assign_grade)
        most_common_grade = df['Programming_skills'].value_counts().idxmax()
        print(df)
        print("Most frequent skill level:", most_common_grade)
        # Create a pie chart
        skill_counts = df['Programming_skills'].value_counts()
        plt.figure(figsize=(8, 8))
        plt.pie(skill counts, labels=skill counts.index, autopct='%1.1f%%', startangle=1
        plt.title('Distribution of Programming Skills')
        plt.show()
                                  PP BEEE FL FIMS Programming_skills
             S.NO SECTION DV M-II
              1.0 ALPHA 12 0 17.0 9 19
       0
                                                 15
                                                              Average
       1
              2.0 ALPHA 19 12 16.0 16 18
                                                 3
                                                              Average
       2
              3.0 ALPHA 18 14 18.0 18 18 16
                                                            Very Good
       3
              4.0 ALPHA 15
                              9 19.0 17 19 15
                                                            Very Good
              5.0 ALPHA 18 17 19.0 19 20 18
                                                            Very Good
              . . .
                    ... .. ... ... ...
       596 597.0 SIGMA 20
                             20 20.0 20 20
                                                20
                                                            Very Good
       597 598.0
                   SIGMA 20
                              20 20.0 19 19
                                                18
                                                            Very Good
       598 599.0 SIGMA 20 20 17.0 17 19 18
                                                              Average
       599 600.0 SIGMA 14 12 11.0 9 18 17
                                                                 Poor
                              19 20.0 18 18 19
       600 601.0
                   SIGMA 20
                                                            Very Good
       [599 rows x 9 columns]
       Most frequent skill level: Poor
```

```
file:///C:/Users/saddi/Downloads/MID MARKS (1).html
```

Distribution of Programming Skills



```
In [55]: subjects = ['DV', 'M-II', 'PP', 'BEEE', 'FL', 'FIMS']
    subset = df[df[subjects].eq(20).any(axis=1)]
    print("Subset of students who scored 20 in any subject:")
    print(subset)
    for subject in subjects:
        count_20 = (df[subject] == 20).sum()
        print(f"Students who scored 20 in {subject}: {count_20}")
```

Subset of students who scored 20 in any subject:

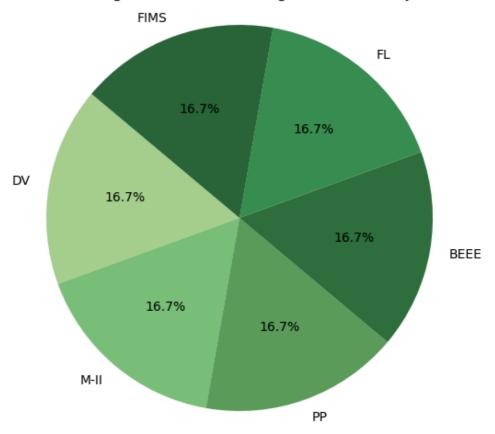
```
PP BEEE FL FIMS Programming_skills
            S.NO SECTION DV M-II
                  ALPHA 18 17 19.0 19 20
       4
             5.0
                                                18
                                                           Very Good
       6
             7.0 ALPHA 15 10 20.0 20 15
                                              14
                                                          Very Good
       7
             8.0 ALPHA 17 17 19.0 20 19 13
                                                          Very Good
             9.0 ALPHA 10 18
                                 0.0 20 19 15
       8
                                                               Poor
       9
            10.0 ALPHA 18 19 20.0 20 20 15
                                                          Very Good
                   ... ... ..
                                                                . . .
            . . .
                            14 16.0 18 20
       595 596.0 SIGMA 17
                                              18
                                                            Average
       596 597.0 SIGMA 20
                            20 20.0 20 20 20
                                                           Very Good
       597 598.0 SIGMA 20 20 20.0 19 19 18
                                                          Very Good
       598 599.0 SIGMA 20 20 17.0 17 19 18
                                                             Average
       600 601.0 SIGMA 20 19 20.0 18 18 19
                                                           Very Good
       [253 rows x 9 columns]
       Students who scored 20 in DV: 88
       Students who scored 20 in M-II: 56
       Students who scored 20 in PP: 104
       Students who scored 20 in BEEE: 89
       Students who scored 20 in FL: 159
       Students who scored 20 in FIMS: 27
In [56]: import pandas as pd
        import matplotlib.pyplot as plt
        data = {
            'DV': [20, 15, 18, 20, 10],
            'M-II': [10, 20, 15, 14, 20],
            'PP': [20, 12, 20, 18, 16],
            'BEEE': [14, 20, 19, 12, 20],
            'FL': [18, 14, 20, 20, 15],
            'FIMS': [10, 20, 16, 14, 20]
        df = pd.DataFrame(data)
        subjects = ['DV', 'M-II', 'PP', 'BEEE', 'FL', 'FIMS']
        counts = [df[subject].eq(20).sum() for subject in subjects]
        green_shades = ['#a8d08d', '#7bbf7b', '#5a9b5a', '#2d703d', '#3b8c50', '#286838'
        plt.figure(figsize=(8, 6))
        plt.pie(counts, labels=subjects, autopct='%1.1f%%', startangle=140, colors=green
```

plt.title('Percentage of Students Scoring 20 in Each Subject')

plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.

plt.show()

Percentage of Students Scoring 20 in Each Subject



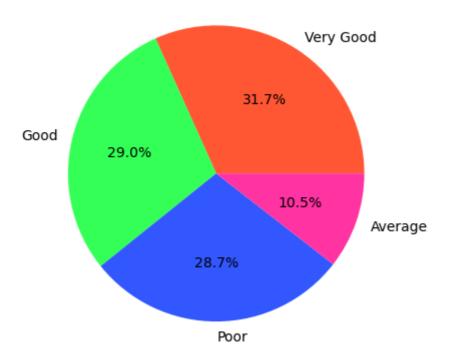
```
In [57]:
         import pandas as pd
          import matplotlib.pyplot as plt
          data = pd.read_excel("MIDMARKS.XLSX")
          subjects = ["DV"]
          data.dropna(inplace=True)
          for subject in subjects:
              data[subject] = pd.to_numeric(data[subject], errors='coerce').fillna(0)
          def assign_grade(percentage):
              if 18 <= percentage <= 20:</pre>
                  return 'Very Good'
              elif 13 <= percentage <= 14:</pre>
                 return 'Average'
              elif 13 <= percentage <= 17:</pre>
                  return 'Good'
              else:
                  return 'Poor'
          data['Programming_skills'] = data['DV'].apply(assign_grade)
          grade_counts = data['Programming_skills'].value_counts()
```

```
custom_colors = ['#FF5733', '#33FF57', '#3357FF', '#FF33A1']

plt.pie(grade_counts, labels=grade_counts.index, autopct='%1.1f%%', colors=custc
plt.title("Distribution of Programming Skills Grades")
plt.show()

print(data)
```

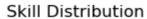
Distribution of Programming Skills Grades

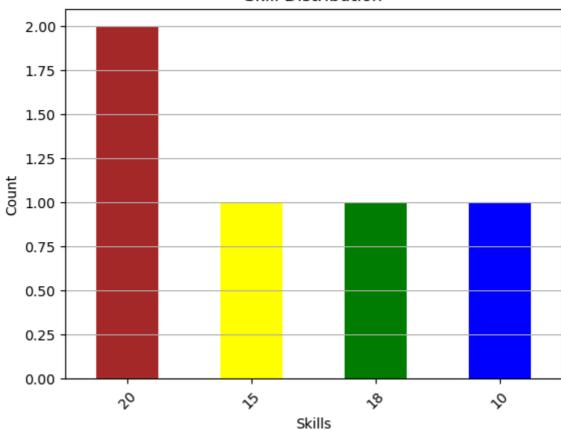


```
S.NO SECTION
                   DV M-II PP BEEE FL FIMS Programming_skills
0
      1.0 ALPHA 12.0
                        0 17
                                9
                                   19
                                       15
                                                      Poor
      2.0
           ALPHA 19.0 12 16
                                                  Very Good
1
                               16 18
                                        3
2
      3.0
           ALPHA 18.0
                       14
                           18
                               18
                                   18
                                       16
                                                  Very Good
3
      4.0
           ALPHA 15.0 9 19
                                   19
                                       15
                                                      Good
                               17
          ALPHA 18.0
                      17 19
                                                  Very Good
      5.0
                              19 20
                                      18
      . . .
            . . .
                  . . .
                      . . .
                           . .
                              . . .
                                       . . .
596 597.0
           SIGMA 20.0 20
                           20
                              20
                                   20
                                      20
                                                  Very Good
   598.0
           SIGMA 20.0 20 20 19 19 18
597
                                                  Very Good
598 599.0
           SIGMA 20.0 20 17
                              17 19 18
                                                  Very Good
           SIGMA 14.0
                               9 18
599
    600.0
                       12 11
                                       17
                                                    Average
600 601.0
           SIGMA 20.0
                       19 20
                                       19
                                                  Very Good
                               18 18
```

[599 rows x 9 columns]

```
import matplotlib.pyplot as plt
skill_counts = df['DV'].value_counts()
skill_counts.plot(kind='bar', color= ["brown" , "yellow" , "green", "blue"])
plt.title('Skill Distribution')
plt.xlabel('Skills')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
print(skill_counts)
```





Name: count, dtype: int64