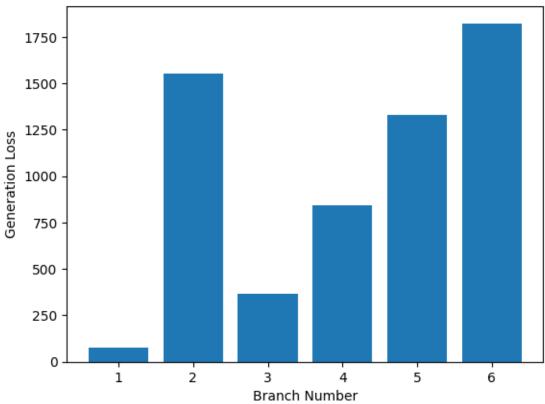
## November 30, 2023

```
[456]: import numpy as np
       import pandas as pd
       from matplotlib import pyplot as plt
       import math
       import itertools
       import seaborn
[457]: #Compute Generation Losses on Loss of Branches
       n=6 #set number of branches here
       gen_loss=np.random.randint(low=30,high=2000,size=(1,n))
       gen_loss=gen_loss[0]
       print(gen_loss)
       list1=[]
       for i in range(1,len(gen_loss)+1):
           list1.append(i)
       print(list1)
       namelist=[]
       for i in list1:
           namelist.append(str(i))
       plt.bar(namelist,gen_loss)
       plt.title('Generation Loss when Branch Goes Offline')
       plt.xlabel('Branch Number')
      plt.ylabel('Generation Loss')
       plt.show()
      [ 77 1555 369 844 1332 1824]
      [1, 2, 3, 4, 5, 6]
```

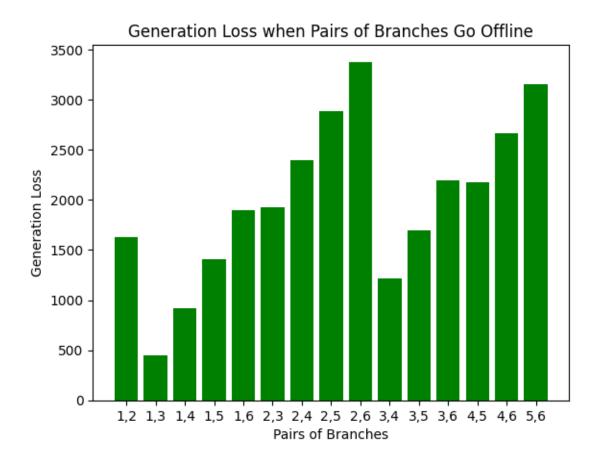




```
[458]: a=math.comb(n,2)
       print(a)
       gen_pair_loss=np.zeros((a),dtype='i')
       print(gen_pair_loss)
       list_of_tuples=[]
       list_of_indices=[]
       i=0
       for comb in itertools.combinations(list1, 2):
           list_of_tuples.append(comb)
           r=str(comb[0])+','+str(comb[1])
           list_of_indices.append(r)
           gen_pair_loss[i]=gen_loss[comb[0]-1]+gen_loss[comb[1]-1]
           i+=1
       print(gen_pair_loss)
       plt.bar(list_of_indices,gen_pair_loss,color='green')
       plt.xlabel("Pairs of Branches")
       plt.ylabel("Generation Loss")
       plt.title("Generation Loss when Pairs of Branches Go Offline")
```

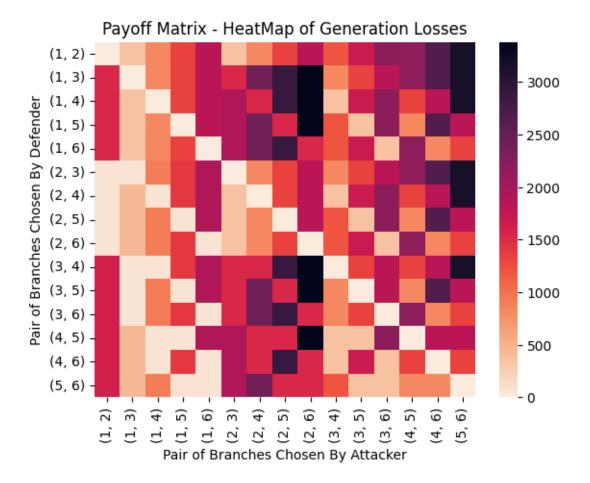
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0] [1632 446 921 1409 1901 1924 2399 2887 3379 1213 1701 2193 2176 2668 3156]

[458]: Text(0.5, 1.0, 'Generation Loss when Pairs of Branches Go Offline')



[[ 0 369 844 1332 1824 369 844 1332 1824 1213 1701 2193 2176 2668 3156]

```
Γ1555
                0 844 1332 1824 1555 2399 2887 3379 844 1332 1824 2176 2668
        3156]
       [1555 369
                     0 1332 1824 1924 1555 2887 3379 369 1701 2193 1332 1824
        3156]
                         0 1824 1924 2399 1555 3379 1213 369 2193 844 2668
       Γ1555
                  844
             369
        18247
                              0 1924 2399 2887 1555 1213 1701 369 2176 844
       [1555
             369
                  844 1332
        13327
       Γ 77
                  921 1409 1901
                                   0 844 1332 1824 844 1332 1824 2176 2668
              77
        31567
       [ 77 446
                   77 1409 1901 369
                                        0 1332 1824 369 1701 2193 1332 1824
        3156]
       [ 77
                  921
                        77 1901
                                 369
                                      844
                                             0 1824 1213 369 2193 844 2668
             446
        1824]
       [ 77 446
                                                  0 1213 1701 369 2176 844
                  921 1409
                             77 369 844 1332
        1332]
       [1632
              77
                   77 1409 1901 1555 1555 2887 3379
                                                       0 1332 1824 1332 1824
        3156]
       [1632
              77 921
                        77 1901 1555 2399 1555 3379 844
                                                            0 1824 844 2668
        18247
       Γ1632
                             77 1555 2399 2887 1555
                  921 1409
                                                     844 1332
                                                                 0 2176
        1332]
       Γ1632 446
                        77 1901 1924 1555 1555 3379
                                                     369
                                                          369 2193
                                                                      0 1824
        18247
       Γ1632 446
                   77 1409
                             77 1924 1555 2887 1555
                                                    369 1701
                                                               369 1332
        1332]
       [1632 446 921
                             77 1924 2399 1555 1555 1213 369 369 844
                        77
                                                                        844
           0]]
[469]: payoff_map=pd.
       DataFrame(payoff,columns=list_of_tuples,index=list_of_tuples,dtype='f')
      seaborn.heatmap(payoff_map,cmap=seaborn.cm.rocket_r)
      plt.title("Payoff Matrix - HeatMap of Generation Losses")
      plt.xlabel("Pair of Branches Chosen By Attacker")
      plt.ylabel("Pair of Branches Chosen By Defender")
      plt.show()
```



```
[476]: #Case-1: If Defender was static::defender choses each defence vector pair with

dequal probability

#then our attacker must chose the column of payoff matrix that has highest sum

dof values

game_value=payoff_map.sum(axis=0).values

for i in range(0,len(game_value)):

    game_value[i]/=math.comb(n,2)

plt.xlabel("Attack Vector")

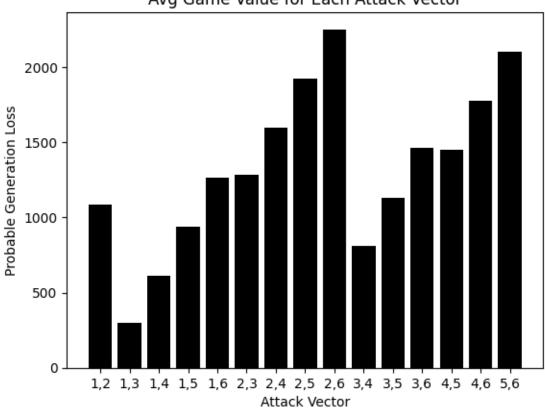
plt.ylabel("Probable Generation Loss")

plt.title("Avg Game Value for Each Attack Vector")

plt.bar(list_of_indices,game_value,color='#0000000')

plt.show()
```





```
[477]: #Case-1 solution
print("Best Attack Vector in Case-1 is: ",payoff_map.sum(axis=0).idxmax())
print("Best Probable Generation Loss is: ",round((payoff_map.

→sum(axis=0)[payoff_map.sum(axis=0).idxmax()])/math.comb(n,2)))
```

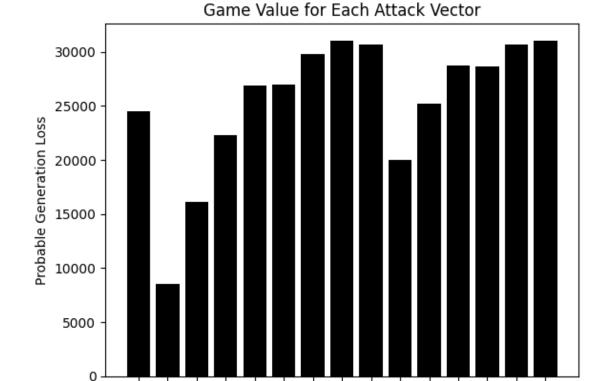
Best Attack Vector in Case-1 is: (2, 6) Best Probable Generation Loss is: 2253

[1.5034441 1.9078078 1.7458576 1.5794749 1.4117289 1.4038874 1.241937

## 1.0755545 0.9078078 1.6463008 1.4799182 1.3121722 1.3179679 1.150222 0.9838393]

```
[479]: #for every attack-vector column, the probable generation-loss is now the mean of modified values. The modified values refer to #the rows of payoff matrix multilplied by corresponding probability factor of payoff-payoff-map.mul(p_vector,axis=1)
```

```
[480]: gr=modified_payoff.sum(axis=0).values
  plt.xlabel("Attack Vector")
  plt.ylabel("Probable Generation Loss")
  plt.title("Game Value for Each Attack Vector")
  plt.bar(list_of_indices,gr,color='#000000')
  plt.show()
```



1,2 1,3 1,4 1,5 1,6 2,3 2,4 2,5 2,6 3,4 3,5 3,6 4,5 4,6 5,6 Attack Vector

```
[481]: #Case-2 solution

print("Best Attack Vector in Case-2 is: ",modified_payoff.sum(axis=0).idxmax())

print("Best Probable Generation Loss is: ",round((payoff_map.

sum(axis=0) [modified_payoff.sum(axis=0).idxmax()])/math.comb(n,2)))
```

Best Attack Vector in Case-2 is: (2, 5) Best Probable Generation Loss is: 1925

[]:[