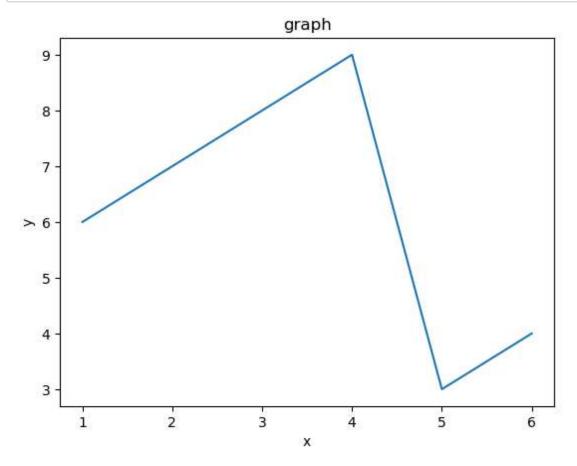
```
In [2]:
        #1)
        import numpy as np
        a=np.array([[1,2,3,],[4,5,6]])
        print('array type is :',type(a))
        print("shape of array:",a.shape)
        print('size of array:',a.size)
        print('array elements type :',a.dtype)
        print('dimension of array:',a.ndim)
        print(a[1,...])
        print('\n')
        print(a[...,1])
        print('\n')
        print(a[...,1:])
        print('\n')
        print(a[1:])
        array type is : <class 'numpy.ndarray'>
        shape of array: (2, 3)
        size of array: 6
        array elements type : int32
        dimension of array: 2
        [4 5 6]
        [2 5]
        [[2 3]
         [5 6]]
        [[4 5 6]]
```

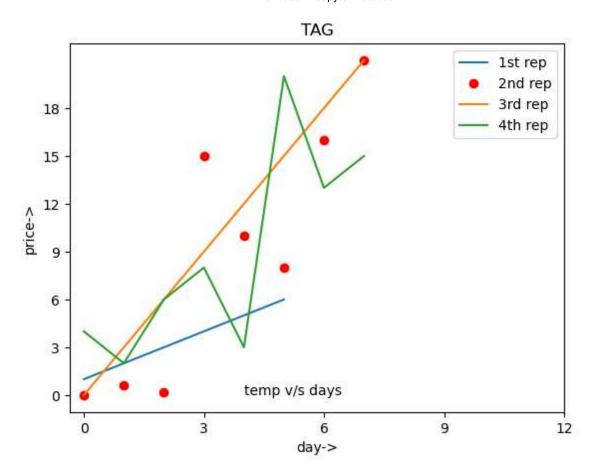
```
In [24]:
         #2)
         import pandas as pd
         import numpy as np
         data=[1,2,3,4,5,6]
         df=pd.DataFrame(data,columns=['numbers'])
         print(df,'\n')
         data1=[[1,'kumar'],[2,'raam'],[3,'soma']]
         sp=pd.DataFrame(data1,columns=['id','names'])
         print(sp,'\n')
         data2={
              'name': ['kumar','raam','bob'],
              'age': [23,24,21],
             'city': ['newyork','bangalore','bhadra']
         }
         rt=pd.DataFrame(data2)
         print(rt,'\n')
         my_df = pd.DataFrame(data=[4,5,6,7], index=range(0,4), columns=['A'])
         print(pd.DataFrame(my_df))
         print('\n')
         my_dict = {1: ['1', '3'], 2: ['1', '2'], 3: ['2', '4']}
         print(pd.DataFrame(my_dict))
```

```
numbers
0
          1
1
          2
2
          3
3
          4
4
          5
5
          6
   id names
0
    1
       kumar
1
    2
         raam
2
    3
         soma
    name
                      city
           age
            23
0
   kumar
                   newyork
1
    raam
            24
                bangalore
2
     bob
                    bhadra
            21
   Α
0
   4
1
   5
2
   6
3
   7
      2 3
   1
0
   1
          2
      1
```

1 3 2 4

```
In [27]:
         #3)
         import matplotlib.pyplot as plt
         x=[1,2,3,4,5,6]
         y=[6,7,8,9,3,4]
         plt.plot(x,y)
         plt.xlabel('x')
         plt.ylabel('y')
         plt.title('graph')
         plt.show()
         a=[1,2,3,4,5,6]
         b=[0,0.6,0.2,15,10,8,16,21]
         plt.plot(a)
         plt.plot(b,"or")
         plt.plot(list(range(0,22,3)))
         plt.xlabel("day->")
         plt.ylabel("price->")
         plt.title("TAG")
         c=[4, 2, 6, 8, 3, 20, 13, 15]
         plt.plot(c,label="4th rep")
         plt.xticks(list(range(0,15,3)))
         plt.yticks(list(range(0,20,3)))
         plt.legend(['1st rep','2nd rep','3rd rep','4th rep'])
         plt.annotate("temp v/s days",xy=(4,0))
         plt.show()
```



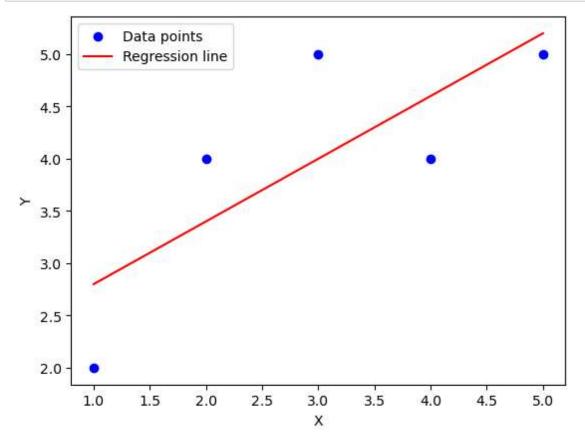


```
In [31]:
         #4)
         import numpy as np
         vector_a=np.array([1,2,3])
         vector_b=np.array([4,5,6])
         addition=np.add(vector_a, vector_b)
         subtraction=np.subtract(vector a, vector b)
         multiply=np.dot(vector_b,vector_a)
         mean_a=np.mean(vector_a)
         magnitude=np.linalg.norm(vector_a)
         distance=np.linalg.norm(vector_a-vector_b)
         sumofsquares=np.sum(np.square(vector_a))
         print(addition,'\n')
         print(subtraction,'\n')
         print(multiply,'\n')
         print(mean_a,'\n')
         print(magnitude,'\n')
         print(distance,'\n')
         print(sumofsquares)
         [5 7 9]
         [-3 -3 -3]
         32
         2.0
         3.7416573867739413
         5.196152422706632
         14
In [34]:
         #5)
         import pandas as pd
         data={
              'x':[1,2,3,4,5],
              'y':[2,2,3,4,5]
         df=pd.DataFrame(data)
         covriance_matrix=df.cov()
         covariance=covriance_matrix.loc['x','y']
         correlation_matrix=df.corr()
         correlation=correlation matrix.loc['x','y']
         print(f" covariance:{covariance}")
         print(f" correlation:{correlation}")
          covariance:2.0
```

correlation:0.9701425001453321

```
In [35]:
          #7
          def gd(sp,lp,ni):
              def sq(x):
                  return x**2
              x=sp
              for i in range(ni):
                  grad=2*x
                  x=x-lp*grad
                  print(f"iteration{i+1}:x=\{x\},f(x)=\{x**2\}")
          sp=10
          lp=0.1
          ni=10
          x_min=gd(sp,lp,ni)
          print(f"optimize at x=\{x \text{ min}\}, f(x)=\{x \text{ min**2}\}")
          print(f'\setminus nAfter \{ni\} iterations, the optimized value is x = \{x min\}')
          iteration1:x=8.0, f(x)=64.0
          iteration2:x=6.4, f(x)=40.96000000000001
          iteration3:x=5.12, f(x)=26.2144
          iteration4:x=4.096, f(x)=16.777216
          iteration5:x=3.2768, f(x)=10.73741824
          iteration6:x=2.62144, f(x)=6.871947673600001
          iteration7:x=2.0971520000000003, f(x)=4.398046511104002
          iteration8:x=1.6777216000000004, f(x)=2.8147497671065613
          iteration9:x=1.3421772800000003, f(x)=1.801439850948199
          iteration10: x=1.0737418240000003, f(x)=1.1529215046068475
          optimize at x=1.0737418240000003, f(x)=1.1529215046068475
          After 10 iterations, the optimized value is x = 1.0737418240000003
```

```
In [40]:
         #8
         import numpy as np
         import matplotlib.pyplot as plt
         def simple linear regression(x, y):
             x mean = np.mean(x)
             y_mean = np.mean(y)
             b1 = np.sum((x - x_mean) * (y - y_mean)) / np.sum((x - x_mean)**2)
             b0 = y_mean - b1 * x_mean
             return b0, b1
         x = np.array([1, 2, 3, 4, 5])
         y = np.array([2, 4, 5, 4, 5])
         b0, b1 = simple_linear_regression(x, y)
         y pred = b0 + b1 * x
         plt.scatter(x, y, color='blue', label='Data points')
         plt.plot(x, y_pred, color='red', label='Regression line')
         plt.xlabel('X')
         plt.ylabel('Y')
         plt.legend()
         plt.show()
         print(f"Linear Regression Line: y = {b0:.2f} + {b1:.2f}x")
```



Linear Regression Line: y = 2.20 + 0.60x

```
In [39]: #9
         def accuracy(tp, fp, fn, tn):
             correct = tp + tn
             total = tp + fp + fn + tn
             return correct / total
         def precision(tp, fp, fn, tn):
             return tp / (tp + fp)
         def recall(tp, fp, fn, tn):
             return tp / (tp + fn)
         def f1_score(tp, fp, fn, tn):
             p = precision(tp, fp, fn, tn)
             r = recall(tp, fp, fn, tn)
             return 2 * p * r / (p + r)
         def evaluate_classification(tp, fp, fn, tn):
             acc = accuracy(tp, fp, fn, tn)
             prec = precision(tp, fp, fn, tn)
             rec = recall(tp, fp, fn, tn)
             f1 = f1_score(tp, fp, fn, tn)
             print(f"Accuracy: {acc:.2f}")
             print(f"Precision: {prec:.2f}")
             print(f"Recall: {rec:.2f}")
             print(f"F1 Score: {f1:.2f}")
         # Take user input for the confusion matrix values
         tp = int(input("Enter True Positive (tp): "))
         fp = int(input("Enter False Positive (fp): "))
         fn = int(input("Enter False Negative (fn): "))
         tn = int(input("Enter True Negative (tn): "))
         # Evaluate the classification performance
         evaluate classification(tp, fp, fn, tn)
```

```
Enter True Positive (tp): 6
Enter False Positive (fp): 9
Enter False Negative (fn): 3
Enter True Negative (tn): 8
Accuracy: 0.54
Precision: 0.40
Recall: 0.67
F1 Score: 0.50
```

```
import random
In [41]:
         def random_kid():
             return random.choice(["boy","girl"])
         both girls=0
         either girl=0
         older_girl=0
         for _ in range(1000):
             younger=random_kid()
             older=random_kid()
             if younger=="girl" and older=="girl":
                 both_girls+=1
             if younger=="girl" or older=="girl":
                 either girl+=1
             if older=="girl":
                 older_girl+=1
         print("the probability of both girls where older is a girl:",both_girls/older_gir
         print("the probability of both girls where one is a girl:",both_girls/either_girl
```

the probability of both girls where older is a girl: 0.4843137254901961 the probability of both girls where one is a girl: 0.3302139037433155

```
In [ ]:
```