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Calculate the mean, median, mode and standard deviation for
          the problem statements 1 & 2.
          Problem Statement 1:
          The marks awarded for an assignment set for a Year 8 class of 20 students were as follows: 6 7 5 7 7 8 7 6 9 7 4 10 6 8 8 9 5
          648
 In [4]: import numpy as np
          from scipy import stats
          import math
          import statistics
          import os
          import sys
 In [5]: | marks = np.asarray([6,7,5,7,7,8,7,6,9,7,4,10,6,8,8,9,5,6,4,8])
 In [6]: def Statistics(m):
              print("Mean :", np.mean(m))
              print("Median :", np.median(m))
              print("Mode :", statistics.mode(m))
              print("Standard Deviation :", np.std(m))
 In [7]: Statistics(marks)
          Mean : 6.85
          Median: 7.0
          Mode: 7
          Standard Deviation: 1.5898113095584647
          Problem Statement 2:
          The number of calls from motorists per day for roadside service was recorded for a particular month:
          28, 122, 217, 130, 120, 86, 80, 90, 140, 120, 70, 40, 145, 113, 90, 68, 174, 194, 170, 100, 75, 104, 97, 75, 123, 100, 75, 104,
          97, 75, 123, 100, 89, 120, 109
 In [8]: CallRec = np.asarray([28, 122, 217, 130, 120, 86, 80, 90, 140, 120, 70, 40, 145, 113, 90, 68
          , 174, 194, 170, 100, 75, 104, 97, 75, 123, 100, 75, 104, 97, 75, 123, 100, 89, 120, 109])
 In [9]: Statistics(CallRec)
          Mean: 107.51428571428572
          Median : 100.0
          Mode : 75
          Standard Deviation : 38.77287080168403
          Problem Statement 3:
          The number of times I go to the gym in weekdays, are given below along with its associated probability: x = 0, 1, 2, 3, 4, 5 f(x)
          = 0.09, 0.15, 0.40, 0.25, 0.10, 0.01 Calculate the mean no. of workouts in a week. Also evaluate the variance involved in it.
In [10]: x = np.asarray([0,1,2,3,4,5])
          f_x = np.array([0.09, 0.15, 0.40, 0.25, 0.10, 0.01])
          print(x.reshape((1,-1)))
          print(f_x.reshape((-1,1)))
          [[0 1 2 3 4 5]]
          [[0.09]
           [0.15]
           [0.4]
           [0.25]
           [0.1]
           [0.01]]
In [11]: x.reshape((1,-1))
          f_x.reshape((-1,1))
          mean=np.dot(x,f_x)
          variance_of_x=(x-mean)**2
          variance = np.dot(variance_of_x.reshape(1,-1),f_x)
          print("Mean no. of workouts:", mean)
          print("Variance of workouts:", variance)
          Mean no. of workouts: 2.15
          Variance of workouts: [1.2275]
          Problem Statement 4:
          Let the continuous random variable D denote the diameter of the hole drilled in an aluminum sheet. The target diameter to be
          achieved is 12.5mm. Random disturbances in the process often result in inaccuracy. Historical data shows that the distribution
          of D can be modelled by the PDF (d) = 20e-20(d-12.5), d \ge 12.5. If a part with diameter > 12.6 mm needs to be scrapped,
          what is the proportion of those parts? What is the CDF when the diameter is of 11 mm? What is your conclusion regarding the
          proportion of scraps?
In [12]: from scipy import integrate
In [13]: PDF=lambda d:20*(np.exp((-20*(d-12.5))))
          x = 12.6
          P_x=integrate.quad(PDF, 12.6, np.inf)
          y = 11
          CDF=integrate.quad(PDF, -np.inf, y)
          print("Proportion of Parts need to scrapped when d >12.6mm is :",P_x[0])
          print("CDF when diameter is 11mm :",CDF[0])
          print("Proportion of CDF when d>12.5mm is :",integrate.quad(PDF,12.5,np.inf)[0])
          Proportion of Parts need to scrapped when d >12.6mm is : 0.13533528323661398
          CDF when diameter is 11mm : nan
          Proportion of CDF when d>12.5mm is : 1.0000000000000024
          C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:1: RuntimeWarning: overflow
          encountered in exp
            """Entry point for launching an IPython kernel.
          C:\ProgramData\Anaconda3\lib\site-packages\scipy\integrate\quadpack.py:364: IntegrationWarnin
          g: The maximum number of subdivisions (50) has been achieved.
            If increasing the limit yields no improvement it is advised to analyze
            the integrand in order to determine the difficulties. If the position of a
            local difficulty can be determined (singularity, discontinuity) one will
            probably gain from splitting up the interval and calling the integrator
            on the subranges. Perhaps a special-purpose integrator should be used.
            warnings.warn(msg, IntegrationWarning)
          Conclusion:
          The conclusion is that function is only valid when d>=12.5.
          When d<12.5, the part can be reworked to 12.5 so no scrap in this case.
          PDF is not defined for d=11
          Problem Statement 5:
          A company manufactures LED bulbs with a faulty rate of 30%. If I randomly select 6 chosen LEDs, what is the probability of
          having 2 faulty LEDs in my sample? Calculate the average value of this process. Also evaluate the standard deviation
          associated with it.
          P = 0.3
          Q = 1 - P = 0.7
          n = total number of trials = 6
          k = number of trail that will be successed = 2
          size = Total number of random samples = 1000
In [14]: from scipy.stats import binom
          import matplotlib.pyplot as plt
          import seaborn as sns
          import math
          binomial_data=binom.rvs(n=6,p=0.3,size=1000)
          sns.distplot(binomial_data, hist=True, kde=True, color="red")
Out[14]: <matplotlib.axes._subplots.AxesSubplot at 0x181e34b1f28>
In [15]: #Probability of getting faulty out of 6 trials
          probab=binom.pmf(k=2, n=6, p=0.3)
          print("Probability of having 2 faulty LED is :",probab)
          cdf=binom.cdf(k=2, n=6, p=0.3)
          print("CDF will be :",cdf)
          Probability of having 2 faulty LED is : 0.3241349999999995
          CDF will be : 0.74431
In [16]: mean, var=binom.stats(n=6, p=0.3)
          print("mean := ", mean)
          print("standard deviation :=", math.sqrt(var))
          mean := 1.799999999999998
          standard deviation := 1.1224972160321822
          Problem Statement 6:
          Gaurav and Barakha are both preparing for entrance exams. Gaurav attempts to solve 8 questions per day with a correction
          rate of 75%, while Barakha averages around 12 questions per day with a correction rate of 45%. What is the probability that
          each of them will solve 5 questions correctly?
          What happens in cases of 4 and 6 correct solutions?
          What do you infer from it? What are the two main governing factors affecting their ability to solve questions correctly? Give a
          pictorial representation of the same to validate your answer.
In [17]: #Gaurav- avg=, p1=0.75
          #Barakha- avg=5, p2=0.45
          #here both the students are independent from each other, correction rate of one doesnot effe
          ct another one
          #G(5)*B(5)
          from scipy.stats import binom
          import numpy as np
          print(f"Probability of each of them solving 5 questions correctly is:{binom.pmf(5,8,0.75)*bi
          nom.pmf(5,12,0.45)}")
          print(f"Probability of each of them solving 4,6 questions correctly is:{binom.pmf(4,8,0.75)*
          binom.pmf(6,12,0.45)")
          #their correction rates effect their combined probability
          Probability of each of them solving 5 questions correctly is:0.04619989057299213
          Probability of each of them solving 4,6 questions correctly is:0.018374956477894576
In [18]: #following graphs show their correction rates invidually and combined
          def binom_plot(n,p,):
              fig, ax=plt.subplots(1,1)
              x = np.arange(binom.ppf(0.01, n, p), binom.ppf(0.99, n, p))
              ax.plot(x, binom.pmf(x, n, p), 'bo', ms=8, label='binom pmf')
              ax.vlines(x, 0, binom.pmf(x, n, p), colors='b', lw=5, alpha=0.5)
In [19]: #Gaurav
          binom_plot(8, 0.75)
           0.30
           0.25
           0.20
           0.15
           0.10
           0.05
           0.00
                3.0
                     3.5
                         4.0
                              4.5
                                        5.5
                                                  6.5
                                    5.0
                                             6.0
In [20]: #Barakha
          binom_plot(12, 0.45)
           0.20
           0.15
           0.10
           0.05
           0.00
In [21]: fig,ax=plt.subplots(1,1)
          x = np.arange(1,11)
          ax.plot(x, binom.pmf(x, 8, 0.75)*binom.pmf(x, 12, 0.45), 'bo', ms=8, label='binom.pmf')
          ax.vlines(x, 0, binom.pmf(x,8,0.75)*binom.pmf(x,12,0.45), colors=\frac{b}{b}, lw=5, alpha=0.5)
          #maximum combined probability observed at 6 question
Out[21]: <matplotlib.collections.LineCollection at 0x181e48f2f60>
           0.06
           0.05
           0.04
           0.03
           0.02
           0.01
           0.00
          Problem Statement 7:
          Customers arrive at a rate of 72 per hour to my shop.
          What is the probability of k customers arriving in 4 minutes? a) 5 customers, b) not more than 3 customers, c) more than 3
          customers.
          Give a pictorial representation of the same to validate your answer.
In [22]: from scipy.stats import poisson
          #We need to calculate average number of customers arriving per 4 minutes
          #72/60 customers come per minute
          mu = 4*(72/60) #customers come per 4 minutes
          print("The probability of arriving 5 cutomers in 4 minutes is :",poisson.pmf(k=5,mu=mu))
          print("The probability of arriving not more than 3 customers in 4 minutes is :",poisson.pmf(
          print("The Probability of more than 3 customers arriving in 4 minutes is : ", 1-poisson.cdf(
          k=3, mu=mu))
          The probability of arriving 5 cutomers in 4 minutes is: 0.17474768364388296
          The probability of arriving not more than 3 customers in 4 minutes is : 0.15169069760753714
          The Probability of more than 3 customers arriving in 4 minutes is : 0.7057700835034357
In [23]: x = list(range(0, 10))
          fig, ax = plt.subplots(1,1,figsize=(15,5))
          ax.plot(x, poisson.pmf(x,mu), 'bo', ms=8, label='poisson pmf')
          ax.vlines(x, 0, poisson.pmf(x, mu), colors='b', lw=5, alpha=0.5)
          plt.xlabel('Number of customers')
          plt.ylabel('Probability')
Out[23]: Text(0,0.5, 'Probability')
            0.175
             0.150
             0.125
             0.100
            0.075
             0.050
             0.025
             0.000
                                                          Number of customers
          Problem Statement 8:
          I work as a data analyst in Aeon Learning Pvt. Ltd. After analyzing data, I make reports, where I have the efficiency of entering
          77 words per minute with 6 errors per hour. What is the probability that I will commit 2 errors in a 455-word financial report?
          What happens when the no. of words increases/decreases (in case of 1000 words,255 words)?
          How is the \lambda affected?
          How does it influence the PMF?
          Give a pictorial representation of the same to validate your answer
In [24]: from scipy.stats import poisson
          #Rate of entering=77 per minute
          #error rate= 6/hour=0.1 per minute
          #No of errors per word=0.1/77
          unit_mu=0.1/77
          def mu(n):
              return n * unit_mu
          print(f"The pobability of committing 2 errors in 455 words financial report is :{poisson.pmf
          (2, mu=mu(455))}")
          print(f"The pobability of committing 2 errors in 1000 words financial report is :{poisson.pmf
          (2, mu=mu(1000))}")
          print(f"The pobability of committing 2 errors in 255 words financial report is :{poisson.pmf
          (2, mu=mu(255))}")
          x=range(100,1000,50)
          mu=[i*unit_mu for i in x]
          fig, ax = plt.subplots(1,1,figsize=(15,5))
          ax.plot(x,poisson.pmf(2,mu), 'bo', ms=8, label='poisson pmf')
          ax.vlines(x,0, poisson.pmf(2,mu), colors='b', lw=5, alpha=0.5)
          #As the number of words increase probability of getting errors increases
          The pobability of committing 2 errors in 455 words financial report is :0.09669027375144444
          The pobability of committing 2 errors in 1000 words financial report is :0.23012815007300153
          The pobability of committing 2 errors in 255 words financial report is :0.039377135392854104
Out[24]: <matplotlib.collections.LineCollection at 0x181e48acd68>
           0.20
           0.15
           0.10
           0.05
           0.00
In [25]: fig, ax = plt.subplots(1,1,figsize=(15,5))
          ax.plot(x,mu, 'bo', ms=8, label='poisson pmf')
          ax.vlines(x,0,mu, colors='b', lw=5, alpha=0.5)
          #Value of mu keeps on increasing with number of words
Out[25]: <matplotlib.collections.LineCollection at 0x181e33922e8>
           1.2
           1.0
           0.8
           0.6
           0.4
           0.2
           0.0
          Problem Statement 10:
          Please compute the following:
          a) P(Z > 1.26), P(Z < -0.86), P(Z > -1.37), P(-1.25 < Z < 0.37), P(Z \le -4.6)
          b) Find the value z such that P(Z > z) = 0.05
          c) Find the value of z such that P(-z < Z < z) = 0.99
In [29]: from scipy.stats import norm
          def P(z,b=-np.inf):
              return integrate.quad(norm.pdf,b,z)[0]
          print('P(Z>1.26) = \%.5f'\%(1-P(1.26)))
          print('P(Z<-0.86) = \%.5f'\%P(-0.86))
          print('P(Z > -1.37) = %.5f'%(1-P(-1.37)))
          print('P(-1.25 < Z < 0.37) = %.5f'%P(0.37, b=-1.25))
          print('P(Z \le -4.6) = \%.5f'\%P(-4.6))
          P(Z>1.26) = 0.89617
          P(Z<-0.86) = 0.19489
          P(Z>-1.37) = 0.91466
          P(-1.25 < Z < 0.37) = 0.53866
          P(Z \le -4.6) = 0.00000
In [28]: print('P(Z>z)=0.05 \text{ is } \%.2f'\%(-1*norm.ppf(0.05)))
          print('P(-z < Z < z) = 0.99 is %.2f'%(abs(norm.ppf(0.005))))
          P(Z>z)=0.05 is 1.64
          P(-z < Z < z) = 0.99 is 2.58
          Problem Statement 11:
          The current flow in a copper wire follow a normal distribution with a mean of 10 mA and a variance of 4 (mA)2.
          What is the probability that a current measurement will exceed 13 mA?
          What is the probability that a current measurement is between 9 and 11mA?
          Determine thecurrent measurement which has a probability of 0.98.
In [30]: mean = 10
          std = np.sqrt(4)
          def I(z, b=-np.inf):
              z = (z-mean)/std
              return integrate.quad(norm.pdf,b,z)[0]
          print(f"Probability that current > 13mA is: {1-I(13)}")
          print(f"Probability that current is between 9 mA and 11 mA is : {1-I(11,b=9)}")
          Probability that current > 13mA is: 0.06680720126885797
          Probability that current is between 9 mA and 11 mA is : 1.3085375387259144
          Problem Statement 12:
          The shaft in a piston has its diameter normally distributed with a mean of 0.2508 inch and a standard deviation of 0.0005 inch.
          The specifications of the shaft are 0.2500 \mp 0.0015 inch.
          What proportion of shafts are in sync with the specifications? If the process is centered so that the mean is equal to the target
          value of 0.2500, what proportion of shafts conform to the new specifications?
          What is your conclusion from this experiment?
```

In [32]: mean\_dia=0.2508

std\_dia=0.0005

#case-1 if mean\_dia=0.2508

8, I(0.2508, 0.0005, 0.2485, 0.2515)}")

0, I(0.2500, 0.0005, 0.2485, 0.2515)}")

def I(mean, std, a, b) :
 a=(a-mean)/std
 b=(b-mean)/std

of regired safts obtained.

#specified dia in the range of 0.2485<d<0.2515

required shafts, there by reducing amount of scrap and reprocessing time.

print(f"Proportion of shafts with dia in range of 0.2485<d<0.2515 when mean diameter:{0.250</pre>

print(f"Proportion of shafts with dia in range of 0.2485<d<0.2515 when mean diameter:{0.250</pre>

Proportion of shafts with dia in range of 0.2485<d<0.2515 when mean diameter:(0.2508, None) Proportion of shafts with dia in range of 0.2485<d<0.2515 when mean diameter:(0.25, None)

When compared to any other manufacturing process whose mean deviates from that of 0.25 less proportion of required shafts are obtained. The more the manufucaturing process deviaties from 0.25, lesser will be the proportion

Mathematically, in a given range 0.2485<d<0.2515, if there are two noraml distributrions (manufacturing processes) with same standard deviation, more area will be covered by the distribution whose mean is closer to mean of the

Within the range of 0.2485<d<0.2515 A manufacturing process with mean of 0.25 gives maximum proportion of