

STATISTICS FOR DATA SCIENCE PART - 3

Estimates of Variability:

- Location is just one dimension in summarizing a feature. There are some other techniques for exploring a feature.
- The other technique is measuring variability also known as dispersion.
- This dispersion measures whether the data is clustered or spread out.
- High dispersion means the data is spread out a lot where as low dispersion means the data is clustered highly.

Key terms in estimating variability:

- The key terms used in dispersion measurement are

Deviation: The difference between observed value and estimation of location. Also known as residual, error.

Variance: The sum of squared deviations from the mean divided by $n - 1$ where n is the number of data. Also known as mean-squared-error.

Standard Deviation: The square root of variance. Also known as l2 norm, euclidean-norm.

Mean absolute deviation: The mean of the absolute value of the deviations from the mean. Also known as l1 norm, Manhattan norm.

Median absolute deviation from median: The median of the absolute value of the deviations from the median.

Range: The difference between the largest and the smallest value in a data set.

Order statistics: Metrics based on the data values sorted from smallest to biggest. Also known as ranks.

Percentile: The value such that P percent of the values take on this value or less and $(100-P)$ percent take on this value or more. Also known as quantile.

Inter quartile range: The difference between the 75th percentile and the 25th percentile. Also known as IQR.

Standard Deviation and Related Estimates:

- Differences are the main basis for calculating measures of dispersion.
- Consider the data elements 1,4,4. The mean of these numbers is 3. The deviations from mean are $1-3, 4-3, 4-3$ which are $-2, 1, 1$.
- The sum of these deviations is 0. This means that the data is not having any dispersion which is incorrect.
- Thus the correct way to measure the dispersion is to take the absolute values of the deviations. Then we end up with the deviations 2,1,1 and the sum of deviations is 4 now which provides a better insight.
- Now the mean of the deviations is $(2+1+1)/3$ which is 1.33. This is known as mean absolute deviation.
- The variance is the mean of sum of squares of all deviations from the mean. For the above data the variance is $(2^2+1^2+1^2)/2$ which is 3. Here instead of 3 which is n the sum is divided with 2 which is $n-1$. This is to make the variance and standard deviation unbiased i.e, it doesn't depend on the sample mean calculation. Thus we have $n-1$ degrees of freedom with one constraint that the variance is calculated based on the calculation of mean.
- Standard deviation is the square root of variance and here it is square root of 3 which is 1.72. This is the most used dispersion measurement.
- The variance and standard deviation are not robust. They are sensitive to outliers.
- The robust dispersion estimation is the median absolute deviation from the median.
- The standard deviation is always greater than the mean absolute deviation, which itself is greater than the median absolute deviation

Estimates Based on Percentiles:

- Sorted data can also be used to understand the dispersion of the data. Statistics based on this technique are called as ordered statistics.
- Range is a good measure to measure dispersion but it is not reliable as it is sensitive to outliers.
- To overcome this percentile concept is used.
- P th percentile is a value such that at least P percent of the values take on this value or less and at least $(100 - P)$ percent of the values take on this value or more.
- To find the 70th percentile first sort the data and then take the 70 percent of data from smallest to the largest then the last number will be 70th percentile.
- Median is a special case which is 50th percentile.
- Quantile is similar to percentile. For example 80th percentile can be considered as 0.8 quantile.
- The common measure of dispersion is the Inter Quantile Range which is the difference between the 75th percentile and the 25th percentile.
- The calculation of the quantiles is expensive as it requires sorting of the data. So machine learning and data science use some approximation measure which will give accuracy upto a certain level.

Python Implementation:

- Now It is time for practical implementation. In this section implementation of dispersion measurement in Python is discussed.
- Consider the data frame *Heart.csv*

```
In [2]: data = pd.read_csv('heart.csv')
```

```
In [3]: data.head()
```

Out[3]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1

- The Libraries to be imported are

```
import numpy as np
import pandas as pd
from scipy import stats
import weighted
```

- The code snippet and output are

```
In [8]: print("Variance:",np.var(data['thalach']))
print("Standard Deviation:",np.std(data['thalach']))
print("IQR:",stats.iqr(data['thalach']))
print("Mean Absolute Deviation:",np.mean(data['thalach'].mad()))
print("Median Absolute Deviation:",np.median(abs(data['thalach']-np.median(data['thalach']))))
```

```
Variance: 522.9148994107331
Standard Deviation: 22.86733258188924
IQR: 32.5
Mean Absolute Deviation: 18.484396954546938
Median Absolute Deviation: 15.0
```