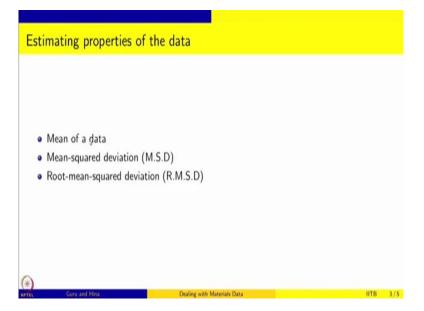
## Dealing with Materials Data Professor M P Gururajan Professor Hina A Gokhale Department of Metallurgical Engineering and Materials Science Indian Institute of Technology, Bombay Lecture 62 Estimating Mean and Mean-Square-Deviation of Data

Welcome to Dealing with Materials Data, we are looking at the collection, analysis and interpretation of data from Material Science and Engineering. And we are in the module on Data Processing.

(Refer Slide Time: 0:26)

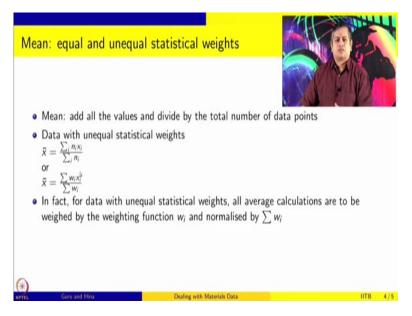
Module 4: Data processing		
<b>F</b> (1) (1) (1) (1) (1) (1) (1) (1) (1) (1)		
Estimating mean and mean-square-deviation of data		
Guru and Hina     Dealing with Materials Data	НТВ	2/5



And in this session we are going to learn about estimating the mean and the mean squared deviation of data. So, these are the properties of data, the mean of the data basically tells there the, if you assume that the data is normally distributed, for example, it tells you actually what is the value about which you see a spread.

But if you do not assume anything about the distribution of the data, the mean also tells you the likelihood estimate or maximum likelihood estimate for the given parameter. Mean-squared deviation basically tells you the spread of the data and root mean-squared deviation is just a root of this quantity. So, it is also a measure of the spread of the data.

(Refer Slide Time: 01:22)

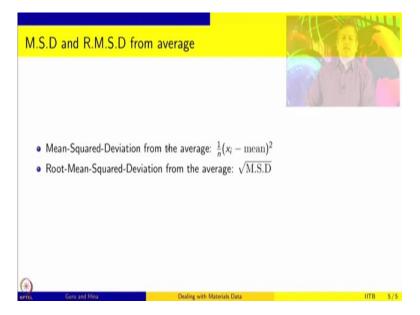


So, mean is very simple. So, you add all values and divide by the total number of data points. So, if you have data with unequal statistical weights, then you have to make sure that you weigh by the weight and then take the average.

$$\bar{x} = \frac{\sum_{i} n_{i} x_{i}}{\sum_{i} n_{i}}$$
or
$$\bar{x} = \frac{\sum w_{i} x_{i}}{\sum w_{i}}$$

And if you take the, these  $w_i$  themselves normalized in such a way that summation  $w_i$  is 1, it is just summation  $w_i$   $x_i$ . We will see an example, we will look at the cluster size frequency data that we have digitized and used to look at taking means of this type.

(Refer Slide Time: 02:21)



Mean squared deviation from average is nothing but, for every value you subtract the mean you square, so it is the mean squared deviation. It is this mean because you then take the mean of those values.

Mean-squared-Deviation from the average:  $\frac{1}{n}(x_i - mean)^2$ 

And root mean squared deviation from the average is just the square root of this value, this is MSD and so RMSD is just square root of this. So let us do this for the data that we have.

## (Refer Slide Time: 02:58)



Let us start R and let us start with this exercise. So I am going to do the DP connectivity data and mean is of course, 101.32, that is straightforward.

## (Refer Slide Time: 3:24)



You can calculate the, okay, so what did we do? We took the conductivity and subtracted the mean from every value, and we squared it, and we took the sum of all the squares and divided by the total number of points, so that was mean squared deviation. And of course, you can also print the root means squared deviation.

So you can see that 0.0096 is the mean squared deviation and the 0.09797959 is the root mean squared deviation.

## (Refer Slide Time: 04:14)

Arthitist Decement viewer *	a Madda	net Dec. 8, 1108 Nata processing using M		1415 V Q 8 000
Out	<pre>msd &lt;- num(y)/length(X\$Conductivity)</pre>			
	<pre>rmsd &lt;- aqrt(msd) print(msd)</pre>			
	## [1] 0.0096			
	print(rmsd)			
	W# [1] 0.09797959			
	$\label{eq:constraint} \begin{array}{l} \mathbf{f} \leftarrow \mathbf{read}, \mbox{case}(^n, . / / Data/ClusterSize) \\ \mathbf{n} \leftarrow \mbox{length}(\mathbf{X} \mathbf{k} \mathbf{x}) \\ \mathbf{f} \mathbf{n} \leftarrow \mathbf{c}(\mathbf{n}, 1) \\ \mathbf{H} \mathbf{D} \leftarrow \mbox{seq}(1, \mathbf{n}, 1) \\ \mathbf{f} \mathbf{or}(\mathbf{i} \mbox{ in } \mathbf{H} \mathbf{D}) \end{array}$	FrequencyZoomed.csv*		
	2[1] = X\$f(1]+X\$s(1] ]] mi = mim(2)/mim(X\$f)			
	print (m)			
	## [1] 239.697			
	af <- aum(X\$f) w <- X\$f/af			
Activities State Vew Parts Session B	und gebug truffe Tunk men	e dec 1, 1109 KStudio		9 45 R HILL 0 01
D a D anna an anna an anna an anna an an an a	C. Martin	A Internet Mary	Catheritiese	E topo dese
- Reality Finite Westman Street		· · · · · · · · · · · · · · · · · · ·	1	C MAL
<pre>&gt; y = c(length(X\$Con</pre>		Values		
and the second	y-mean(X\$Conductivity)	i	9	
> y <- y*y		15044	(5)) a series a series and series.	
<pre>&gt; msd &lt;- sum(y)/leng</pre>	th(X\$Conductivity)	IND	num [1:9] 1 2 3 4 5 6 7	89
> rmsd <- sqrt(msd)		nsd	0.009600000000003	
> print(msd)		mu	239.69696969697	
[1] 0.0096		n	9L	
> print(rmsd)		Files Plata Packages	Neij Vanar	+0
[1] 0.09797959		Plan Jine	wi. 9 6	3 hann -
<pre>&gt; X &lt;- read.csv("Dat</pre>	a/ClusterSizeFrequencyZoomed.csv")			
> n <- length(X\$x)				
> Z <- c(n,1)		8 -		
> IND <- seq(1,n,1)		8 -		
> for(i in IND){		9 -		
+ Z[i] = X\$f[i]*	X\$x[i]		×	
+ }		× 8 -		
> mu = sum(Z)/sum(XS	f)	8 -		
> print(mu)		e -	*	
[1] 239.697		0		
> plot(X\$x,X\$f)			200 250	300
>			XSa	
6		- F.		
🙉 🚍 👩 🗟 🗋	📄 🖾 🔞 🙆 🚨			

And how do we do this for data? Okay, let us do that exercise also with different statistical weights right. So, that is what we want to do. First let us calculate the mean, the average. Okay, so we need to read the data and we need to decide how many data points are there. And then we need to give the weight, the frequency and then we have to divide by the sum of the frequency.

Because this is the w<sub>i</sub>, the frequency is the statistical weight. And we are going to add them all up so that we will normalize it. So, this is the way to calculate the average. So this is 239 and because the data as you have seen. So if you plot, you will see that the peak is somewhere around 240. So, the average turns out to be 240, so that is expected.

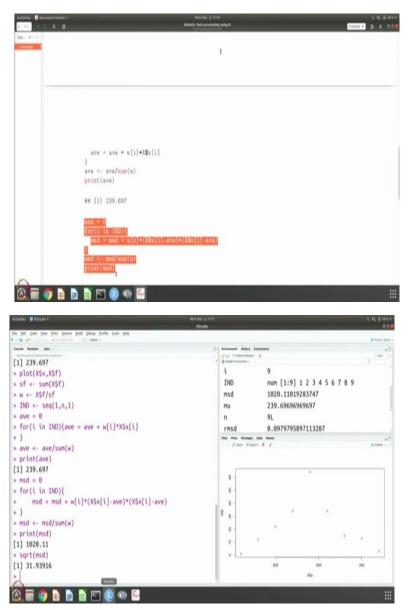
(Refer Slide Time: 05:34)

Addeline Decement Viewer *	Main Dari 2, 1108 Maddel: Data processing entry R	7.41 € mit+
Out., + +	## [1] 0.09797959	
	<pre>X &lt;- read.csv("/,./Data/ClusterSizeFrequencyZoomed.csv") n &lt;- length(X\$x) Z &lt;- c(n,1) IND &lt;- seq(1,n,1) for(i in IRD){ Z[i] = X\$f(i]&gt;X\$x[i] } mu = sum(2)/sum(X\$f) print(mu) ## [1] 239.697  of &lt;- num(X\$i] y &lt;- x\$f(sf[ IDD &lt;- seq(1,n,1) seq = d </pre>	
	for(1 in [107)] :	
	1	
<u>a</u> 💿 🖬 🗈	i 🗈 🖸 🕸 🖗 🖳	
Activities 🗳 Occurrent Viewert *	Mais Dar, L.1100 Makhte (bida percentelity uning M	2 4 8 mm *
Out	1	
	ave - ave + W[1]+X\$x[1] we - ave/sum(v) print(sve)	
	## [1] 239,§97 mad = 0	
	<pre>mod v fib(i n)DD(     mod = nmd + v[i]•(X\$x[i]-ave)•(X\$x[i]-ave) ) mod &lt;- mod/sum(v) prist(mod)</pre>	

	es 1, 1939 Audio		2.4.8.
in 1988 Code Year Ports Service Build Service Ports Service Service			Tran In
<pre>x &lt;- read.csv("Data/ClusterSizeFrequencyZoomed.csv") &gt; A &lt;- length(X\$x) &gt; Z &lt;- c(n, 1)</pre>	A the second states Can A the second states and A the second states MSd MU	0.0096000000000003 239.69696969697	1.00
<pre>&gt; INO &lt;- seq(1,n,1) &gt; INO &lt;- seq(1,n,1) &gt; for(i in INO){ + Z[i] = X\$f[i]*X\$x[i] + }</pre>	n rmsd sf w	9L 0.0979795897113287 198L num [1:9] 0.00505 0.0	16061 0.11111 0
<pre>&gt; mu = sum(Z)/sum(X\$f) &gt; print(mu) [1] 239.697</pre>	Plan Parts Partager III / Jan / Jann -		S Parts -
> plot(X5x,X5f) > sf <- sum(X5f) > w <- X5f/sf > 1N0 <- seq(1,n,1) > ave = 0 > for(i in IND){ave = ave + w[i]*X5x[i]			
+ } > ave <- ave/sum(w) > print(ave) I [1] 239.697 >	g	200 250 X3s	300

Now, let us calculate the mean squared deviation and root mean squared deviation. Okay, so this is another way you can get the weights themselves normalized first, so that they add up to 1 and of course you will get the same number because it is just the algebra, nothing else is different.

(Refer Slide Time: 06:13)



So now let us calculate the mean squared deviation. So how do we calculate the mean squared deviation? You can see that it is the same way. So we take each x value and subtract the average and square the value. But now this has to be weighed by the weighting function. And weighting function is something that we just now calculated.

So we are and remember it is the other type of waiting function, so I took all the frequencies. First I summed all the frequencies, so divided by it. So I have the weighing factor. This weighting factor is what I am going to use. So this is a statistical weight for me to do the calculations. So mean squared deviation is nothing but mean squared deviation and this sum w is going to give 1, so that is the mean squared deviation. So I see mean squared deviation is 1020 and root mean square deviation of course is nothing but the square root of MSD. So, you get about 32 as these spread of this data which is what is given here okay.

So, to summarize, in this session we have seen that you can take data, it could be raw data or it could be data with unequal statistical weights that you got from some analyzed data. In both cases you can calculate the average and you can calculate the spread of the data by looking at how far away from the average the data points lie. And so, we use mean squared deviation and root mean squared deviation to get this spread of the data.

So, we have shown that for 2 cases, one is copper conductivity, the other one is the particle size of titanium aggregates that we have taken from the literature. So we will have more such exercises during these weeks sessions for you to become familiar with this kind of analysis. Thank you.