Predictive Analytics - Regression and Classification Prof. Sourish Das Department of Mathematics Chennai Mathematical Institute

Lecture - 05 Categorical Variable as Predictor Part - 1

Welcome to the Predictive Analytics – Regression and Classification course. In this lecture, we are going to discuss how a Categorical Variable as Predictor is going to play in the linear model setup and typically these models sometimes called ANOVA. And, in a special case it is called ANOVA, but overall it is part of the linear model setups and regression also is part of the linear model setup.

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Experiment

An experiment performed to assess the relative effect of three toxins and a control on the liver of a certain species of trout. The are about the amounts of deterioration (in standard units) of the liver in each sacrificed fish.

Toxin 2	Toxin 3	Control
33	18	11
36	21	14
34	20	11
29	22	16
	Toxin 2 33 36 34 29	Toxin 2 Toxin 3 33 18 36 21 34 20 29 22

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First, we will consider an experiment. So, suppose an experiment is performed to assess the relative effect of three toxins and the control on the liver of a certain species of trout. Trout is

a particular fish a particular species and the effect of toxin on that particular fish that is being studied there are about amounts of deterioration of the in standard units of the liver in each sacrificed fish.

So, you can see there are three toxin levels type of toxin 1, toxin 2 and toxin 3 and then there is a control group where no toxin is being given and the level of deterioration of the liver condition is being studied. So, there are four in each group there are four fishes I mean specimens. So, total there are 16 observation like in each group there are 4. So, there are 4 groups, 3 treatment group and 1 control groups.

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Now, we want to develop a model which will affect the treatment effect. So, what the simple model that we can think of that Y ij i stands for ith group in this case i equal to 1 to k, but k will be for our case is 4 because we have four groups. And, j j runs for within a group how

many samples we have. Here we have in each group we have four samples that, but you can have different number of samples.

So, that is why j runs from 1 to up to n i; n i is the number of samples in the ith group. So, the model that we are thinking is Y ij equal to theta i plus epsilon ij where expectation of epsilon ij is 0 and variance of epsilon ij is sigma i square for where sigma i square finite and covariance of epsilon ij and epsilon i dash j dash is 0.

So, that means, basically we are assuming that each there is no correlation between each of the samples. Each of the samples are independent if you assume that then covariance of epsilon ij and epsilon i dash j dash will be 0.

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So, little bit more clarification of the model. We can take sigma i square to be different we can take sigma i square to be different, but for a simplicity we are assuming sigma i square is equal to sigma square for all groups. So, all group has the similar variance. So, this assumption is called homoscedasticity assumption and then we assume that epsilon ij is following normal distribution with mean at 0 and variance sigma square.

So, this is the model setup that we are going to work out. So, this is the from the model we are framing from the experiment we are framing the model.

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Framing the Model	NPTEL
The above experiment can be modeled as	
$Y_{ij} = \theta_i + \varepsilon_{ij}, i = 1, 2, \cdots, k$ and $j = 1, 2, \cdots, n_i$	
where	
Can we express this model into linear model form?	
$oldsymbol{y} = oldsymbol{X}eta + \epsilon$	
Then we will have an analytic solution to above model.	
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Now, as we frame the model question is can we express this model into linear model form like y equal to x beta plus epsilon? Then we will have an analytic solution of the above model I mean this model will have a analytic solution.

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So, first we have to define the X matrix or the design matrix. So, we have 16 observation for you can see there. So, we have 16 rows here each the first 4 rows of my data belongs to toxin group 1 and rest of the places called toxin 2 for the toxin 2 I created 0s, toxin 3 I created 0 and the control group 0s.

Similarly, the from row 5 to row 8, I have toxin group all 1s; toxin 2 goes to 1; then 3 is 0 and control group from row 5 to row 8 is 0. This is called dummy creation of dummy variable in statistics in machine learning often it is called one hot encoding. This is very useful this one hot encoding is very useful solving many problems and we will see soon how it is going to make our life very simple.

So, we have 16 rows for see each samples we have one representation and if it goes to toxin 1 we have 1, if it not if a particular row does not belong to that group then that row will get 0.

Finding Solutions		()
We have the response vector ${m y}$ as follow	WS:	NETEL
$\mathbf{y} = \begin{bmatrix} 28\\ 23\\ 14\\ 27\\ 33\\ 36\\ 34\\ 29\\ 18\\ 21\\ 20\\ 22\\ 11\\ 14\\ 11\\ 16 \end{bmatrix}$	c ^m i	69
	$\langle \Box \rangle \langle \overline{\sigma} \rangle \langle \overline{z} \rangle \langle \overline{z} \rangle \langle \overline{z} \rangle \langle \overline{z} \rangle$	- P

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All the observations are stacked over another. The first four row observations are belongs to toxin 1; from 5 to 8 observation number 5 to 8 belongs to toxin group 2 from 8 to 12 belongs to toxin group 3 and from 12 13 to 16 it group belongs to control group.

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Then we calculated X transpose X turns out X transpose X in this case become very simple.

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If you go back to the X what happens is X transpose X is essentially it X transpose is kind of you know when it is first row first column will be just it is with itself. So, it will have created 4 and then 0 0 0 in this way the X transpose X creates this matrix. So, X transpose X inverse is very simple, just diagonals will be 1 out of 4 and the half diagonals are all 0. So, this is simple solution.

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Now, we calculate X transpose y. X transpose y is if you look it carefully that sum of the first group of observations the second of element is the sum of the second group of observation and third is the third group of observation and fourth is the fourth group of observation fourth element.

Hence the solution is X transpose X inverse X transpose y. X transpose X inverse is all diagonal elements is 1 by 4 and off diagonals are 0. So, my beta hat is basically sample group mean y 1 bar on the of the first group. Similarly, second. So, y 2 bar is the sample group mean of the second group of the toxin group 2.

y 3 bar stands for the y sample mean for the group 3 in this case toxin 3 and y 4 bar is the sample mean of the control group. So, model yields group means group sample means as the solution for theta 1, theta 2, theta 3 and theta 4.

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Ask yourself: How to model globa	al mean?	NPTEL
The above experiment is be model	led as	
$Y_{ij} = heta_i + arepsilon_{ij}, \ i = 1, 2, \cdots, k$	and $j = 1, 2, \cdots, n_i$	
where		
Each θ _i is the group mean. But how mean of all data?	w can we model a global	
Take a pause of 10 minutes and try the model with global mean.	y for yourself to develop	
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So, this is very interesting solution. Now, I am posing a question to you guys, ask yourself how to model global mean? The above experiment is to be modelled as is being modelled as Y ij equal to theta i plus epsilon ij where i equal to 1 to k and j runs from 1 to n i, where each theta i is the group mean. We realize that each of the theta i is the group mean, but how can we model a global mean for the old all data.

So, we also want to incorporate global mean for all data. In this case situation I would like you to take a pause, pause your video here for about 10 minutes think about it and try for yourself to develop the model with global mean.

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A model with global mean	
► We can try	
$Y_{ij} = \mu \stackrel{\scriptstyle I}{+} \theta_i + \varepsilon_{ij}, i = 1, 2, \cdots, k \text{ and } j = 1, 2, \cdots, n_i$	
where	
But we are going to face a problem. Can you identify the problem?	
Pause the video and think about it for 10 minutes.	
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I believe you have now got a possible solution. Let us see how can you solve the solutions get a solution. If one possible solution is you take Y ij equal to mu mu is the sort of a global mean plus theta i theta i are the group means plus some epsilon ij. Now, i equal to runs to 1 to k and j equal to 1s to n i. This is a simple solution looks like, but we are going to face a problem. Can you identify the problem? So, again I will request you to pause the video for a while and think about maybe for 10 minutes that what problem you are going to face this particular model.

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A model wit Our respon the design	h global se vector j matrix X . I	mean is still s let us se	ame. Th e the cha	e change anged de	e we see esign mat	is in rix.	NPTEL
	/Intercept	Toxin 1	Toxin 2	Toxin 3	Control		
	1	1	0	0	0		
	1	1	0	0	0		
	1	1	0	0	0		
	1	1	0	0	0		
	1	0	1,	0	0		
	1	0	1	0	0		
	1	0	1	0	0		
X =	1	0	1	0	0		
	1	0	0	1	0		
	1	0	0	1	0		
	1	0	0	1	0		
	1	0	0	1	0		
	1	0	0	0	1	1000 C 10	
	1	0	0	0	1	m:	
	1	0	0	0	1	0.1	60
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I believe now you have you have got the idea that what problem you are going to solve you are going to face and let us we know that in this case our response vector will still be same. The change we will see if you look into the model our response vector will going to be same. The only change that we are going to see is in the design matrix.

So, the changes that whatever design matrix we have had before that same design matrix we are still going to have, but now what is what we will have is we are going to have a fifth column called intercept, ok.

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So, we are going to have a intercept column. Now, this is an interesting phenomena that if we have this design matrix then our X transpose X matrix is going to be 16 4 4 4 4 and then 4 4 0 0 0, then 4 0 4 0 0, 4 0 0 4 0, 4 0 0 0 4. In this matrix, if you look at very carefully the first column of this X transpose X is actually direct sum of second column, third column, fourth column and fifth column.

So, if you just add the second column, third column, fourth column and fifth column you will get the first column back. So, the first column is completely dependent on the second, third, fourth and fifth column. So, that means, X transpose X is not going to be invertible; that means, solution does not exist, if we create a dummy variable for each labels of categorical variable. I hope you understand the problem. Let me repeat.

We can let me go back to the model. In this model what we have? We have only effectively one predictor; predictor is treatment and there are four possible levels of the treatment toxin 1, toxin 2, toxin 3 and control. So, for each levels of the treatment so, there is only one predictor that is treatment in treatment level there are four possible levels. So, one categorical variable with four levels – toxin 1, toxin 2, toxin 3 and control group.

For each level if you create a if you create a for each level you create a dummy variable or one hot encoding, then what happens if and also you keep a intercept parameters then the in the X matrix the intercept parameter becomes completely dependent on all the columns of your predictor variables.

As a result one column will become completely dependent on some other columns and hence we will not be going to we are not going to have a invertible matrix or going to have a analytical solution at all. For this case we will solution does not exist. So, you have to be very careful about when you are going to handle categorical variables with different levels.

Let us stop here and we will continue on the next part.