

# MIXED EFFECTS MODELS RANDOM EFFECT

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***"THE FUTURE BELONGS TO THOSE WHO BELIEVE IN THE BEAUTY OF THEIR DREAMS."***

Eleanor Roosevelt

Hello there, readers ! Yes, you. I need your eyes and attention here in 3 ..... 2 ..... 1 .....

Today, what we are interested to discuss is about Random effect in Mixed Effects Models.

Now, the model which we have used so far is of the form :

$$y_{ijk} = \mu + \alpha_i + \beta_j + \epsilon_{ijk} \quad (1)$$

where,  $i = 1, 2$  ;  $j = 1, 2, 3$  ;  $k = 1, 2, \dots, 5$ ; and

$$\epsilon_{ijk} \sim N(0, \sigma_e^2) \quad (2)$$

Here, in this model we are not interested in  $\beta_j$ 's as they are not fixed entities and they just represent  $j$ -th village. As we need only some random villages and not complete enumeration of all villages, so we consider  $\beta_j$ 's as some random quantity, where  $j$  is random and  $\beta$  is the corresponding effect of that randomly chosen village. So, we assume that

$$\beta_j \sim N(0, \sigma_b^2) \quad (3)$$

It means  $\beta_j$  's are i.i.d normal with mean 0 and variance  $\sigma_b^2$  . So, to distinguish between the two different variances, we used  $e$  as a subscript in the earlier variance term. We assume that both of them are independent random variables. Since, it is customary not to use greek letter so instead of  $\beta$  we simply use an english alphabet  $b$ . So, this is our new model :

$$y_{ijk} = \mu + \alpha_i + b_j + \epsilon_{ijk} \quad (4)$$

where,

$$\epsilon_{ijk} \sim N(0, \sigma_e^2) \quad (5)$$

and

$$b_j \sim N(0, \sigma_b^2) \quad (6)$$

Here,  $\alpha_i$  is an unknown parameter and  $b_j$  as well as  $\epsilon_{ijk}$  are random quantities following given normal distributions.

As the villages are chosen randomly, so their effects are also random but the two varieties are not chosen randomly and are of main importance for anyone working on the results. So in this case we are allowing our  $\sigma_b^2$  to be equal to 0 which makes it a special case. It is a kind of degenerate case which conveys the fact that the villages donot have any effect at all. So, the assumption that all  $\beta_j$ 's are same now have all analogous  $b_j$ 's equal to 0, where  $\sigma_b^2$  measures the variability among the villages. We are not interested in specific effect of villages but in the overall effect of the villages.

It is important to know whether villages have any effect or not as then only we can test whether  $\sigma_b^2$  is equal to 0 or not.

In certain models like the one above, these ( $b_j$  and  $\epsilon_{ijk}$ ) are known as **Random Effects** and we call  $\alpha_i$ 's our **Fixed Effects**. A linear model which contains both random as well as fixed effects is termed as **Mixed Effects Model**.

If we have only random effects, then we call the model as **Random Effects Model** and if we only have fixed effects, then we call the model as **Fixed Effects Model**.

That's all for today's session.

We shall continue with why these are useful and the extra computational burden when we are working with such models in our next session.

Thank You for reading with patience.