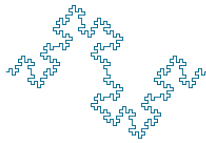


# High-Throughput Sequencing Course

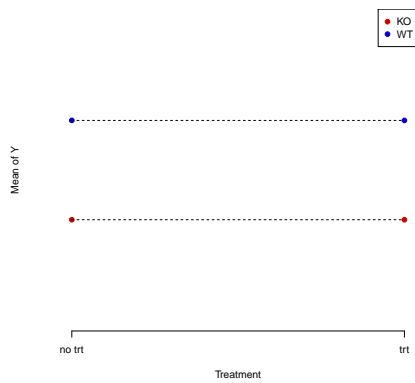
## Statistical Interaction

Biostatistics and Bioinformatics

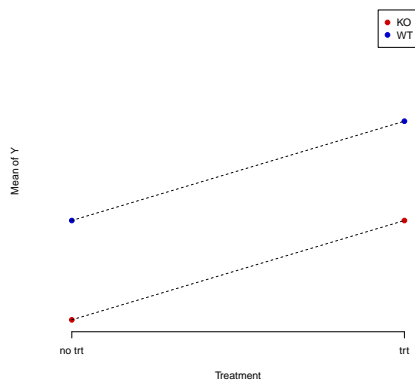


Summer 2019

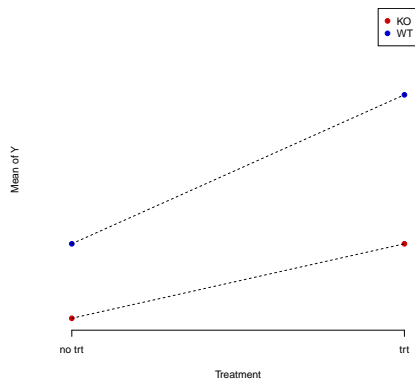
### EXAMPLE 1: NO INTERACTION



### EXAMPLE 2: NO INTERACTION



### EXAMPLE 3: INTERACTION



### MODEL INTERACTION

- $Y$  denotes the gene expression
- Let  $x$  denote the treatment indicator
  - $x = 0$  if not treated or 1 if treated
- Let  $z$  denote the knock-out indicator
  - $z = 0$  is WT or 1 otherwise
- The expected value of  $Y$  given treatment indicator  $x$  and knock out indicator  $z$  is denoted by

$$\mu_{x,z} = E[Y|X = x, Z = z]$$

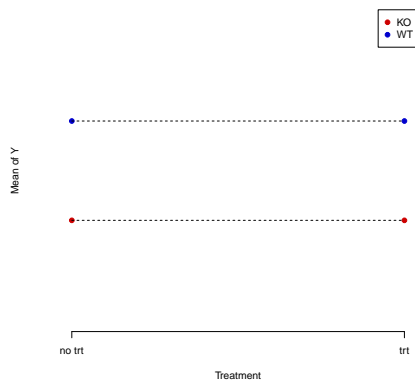
- The model will be

$$Y = \mu_{x,z} + \epsilon$$

where  $\epsilon$  is a the measurement error

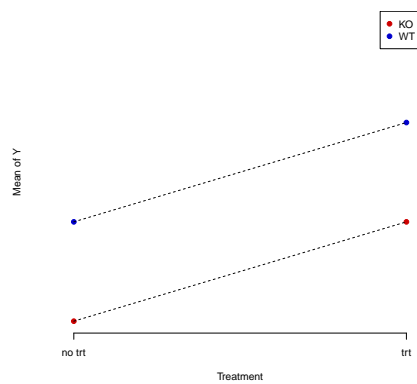
### EXAMPLE 1: LINEAR MODEL FOR NO INTERACTION

$$Y = \beta_0 + \beta_1 z + \epsilon \quad (\beta_2 = 0)$$



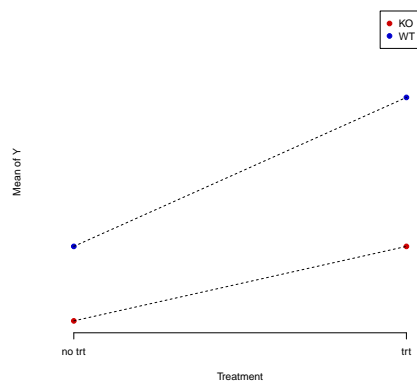
## EXAMPLE 2: NO INTERACTION

$$Y = \beta_0 + \beta_1 z + \beta_2 x + \epsilon$$



## EXAMPLE 3: INTERACTION

$$Y = \beta_0 + \beta_1 z + \beta_2 x + \beta_3 xz + \epsilon$$



## INTERACTION EXAMPLES

- Example 1: What are the signs for  $\beta_0$  and  $\beta_1$ ?
- Example 2: What are the signs for  $\beta_0, \beta_1$  and  $\beta_2$ ?
- Example 2: What are the signs for  $\beta_0, \beta_1, \beta_2$  and  $\beta_3$ ?

## INCORPORATING INTERACTIONS INTO THE NB MODEL

- Conditional on  $x_j$  and  $z_j$ , the observed number of reads mapped to gene  $i$  for sample  $j$ ,  $K_{ij}$ , follows a negative binomial distribution with
  - Mean  $\mu_{ij}$
  - Dispersion parameter  $\alpha_i$  (gene specific)
- Conditional on the treatment status of sample  $j$  ( $x_j = 0$  or  $1$ ) and the temperature  $z_j$ , the expected value of  $K_{ij}$  is

$$\mu_{ij} = s_j \times q_{ij}$$

where

$$\log q_{ij} = \beta_{i0} + \beta_{i1}x_j + \beta_{i2}z_j + \beta_{i3}x_jz_j$$

- The normalization parameters are assumed to be sample (not gene) specific ( $s_{ij} = s_j$ )

## 2019 PILOT EXPERIMENT

- Condition ( $X$ ): pH4 (=0) versus pH8 (=1)
- Strain ( $Z$ ): WT (=0) versus sre1d (=1)
- Negative Binomial Mean Model

$$2^{\beta_0 + \beta_1 X + \beta_2 Z + \beta_3 XZ}$$

- Log 2 of the mean model

$$\beta_0 + \beta_1 X + \beta_2 Z + \beta_3 XZ$$

## 2019 PILOT EXPERIMENT

| Condition | Genotype | Log 2 mean                              |
|-----------|----------|---|
| 0         | 0        | $\beta_0$                               |
| 1         | 0        | $\beta_0 + \beta_1$                     |
| 0         | 1        | $\beta_0 + \beta_2$                     |
| 1         | 1        | $\beta_0 + \beta_1 + \beta_2 + \beta_3$ |

- DE (due to condition) when genotype is WT ( $Z = 0$ ):

$$\begin{aligned} & \{\beta_0 + \beta_1(1) + \beta_2(0) + \beta_3(1)(0)\} \\ & - \{\beta_0 + \beta_1(0) + \beta_2(0) + \beta_3(1)(0)\} \\ & = \beta_1 \end{aligned}$$

- DE (due to treatment) when genotype is sre1d ( $Z = 1$ ):

$$\begin{aligned} & \{\beta_0 + \beta_1(1) + \beta_2(1) + \beta_3(1)(1)\} \\ & - \{\beta_0 + \beta_1(0) + \beta_2(0) + \beta_3(1)(0)\} \\ & = \beta_1 + \beta_2 + \beta_3 \end{aligned}$$

- Note that there is no interaction if  $\beta_3 = 0$

## 2019 PILOT EXPERIMENT

- ▶  $\beta_1$  is log2 fold change (due to condition) when genotype is WT
- ▶  $\beta_1 + \beta_3$  is log2 fold change (due to condition) when genotype is sre1d
- ▶  $2^{\beta_1}$  is the fold change (due to condition) when genotype is WT
- ▶  $2^{\beta_1 + \beta_3}$  is the fold change (due to condition) when genotype is sre1d
- ▶ The fold change of the two fold changes is

$$\frac{2^{\beta_1 + \beta_3}}{2^{\beta_1}} = \frac{2^{\beta_1} 2^{\beta_3}}{2^{\beta_1}} = 2^{\beta_3}.$$

When  $\beta_3 = 0$  we have  $2^0 = 1$