

# Binary predictor, same model

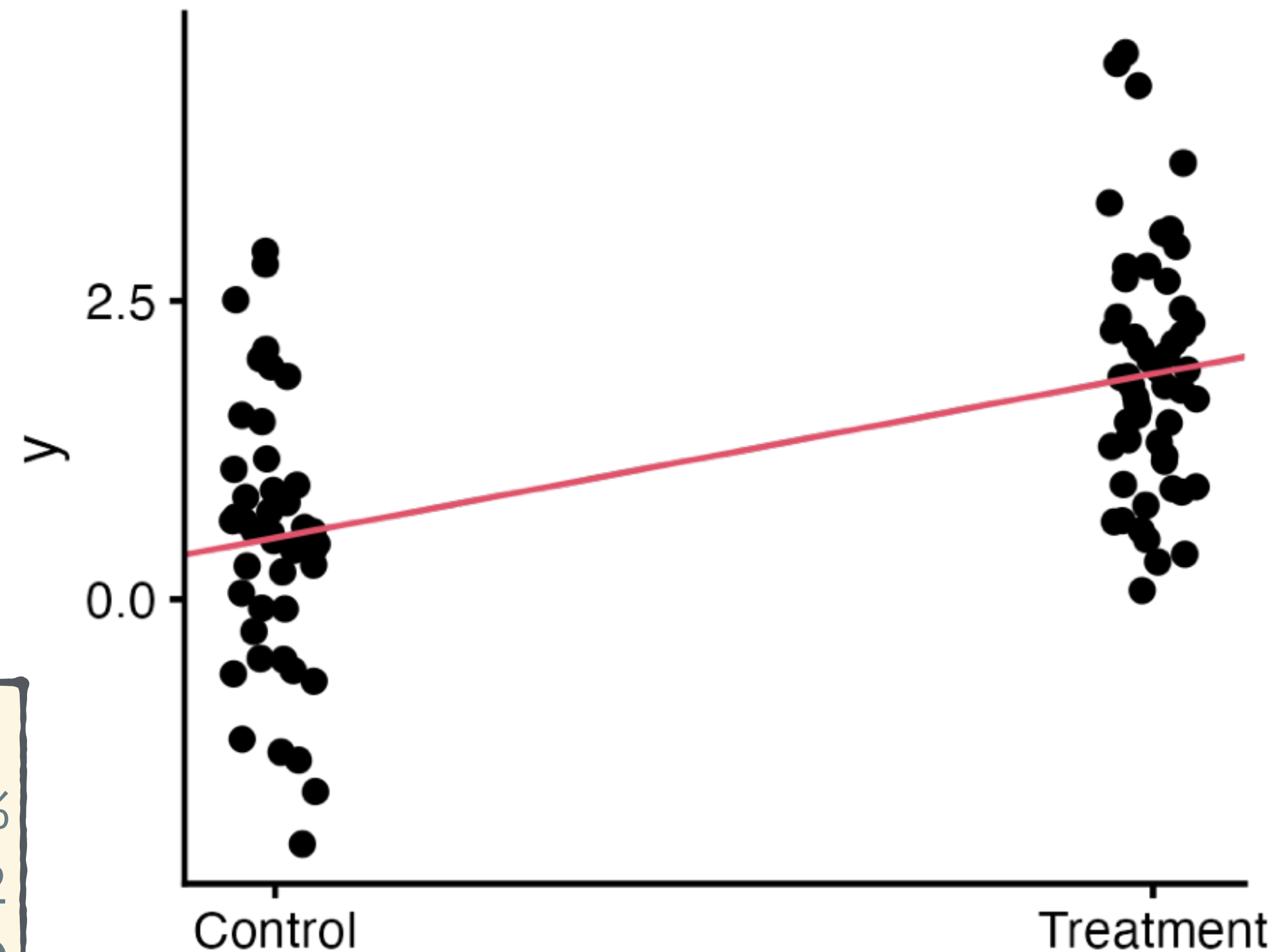
Control-treatment, two categories...

$$y_i \sim N(\mu_i, \sigma)$$

$$\mu_i = \alpha + \beta x_i$$

- $x_i$ : 0 for control, 1 for treatment
- $\alpha$ : the intercept is the **mean of the Control** group
- $\beta$ : the slope is the **difference in the means across the groups**

```
> precis(fit, prob = 0.95)
      mean    sd 2.5% 97.5%
a    0.52 0.16 0.20  0.82
b    1.37 0.22 0.96  1.79
sigma 1.08 0.08 0.94  1.23
```



# How about more categories?

- There are a few ways of modeling predictors with many categories:
  - **Contrasts**: Each category is compared to a **baseline**, and the coefficients are comparisons between baseline and levels
  - **One-hot**: coefficients are means of each level of the predictor
  - **Residuals**: an overall mean is measured, and coefficients are differences between each level and the global mean

```
> x = sample(LETTERS[1:3], 9, replace = TRUE)
> y = 1 + ifelse(x == "A", 0,
                ifelse(x == "B", 1, 2)) + rnorm(9)
> df = tibble(y, x)
> df
# A tibble: 9 × 2
      y x
  <dbl> <chr>
1  1.25 B
2  0.870 A
3 -0.00180 A
4  1.18 B
5  2.03 C
6  2.60 B
7  2.55 B
8  3.92 C
9  4.66 B
```