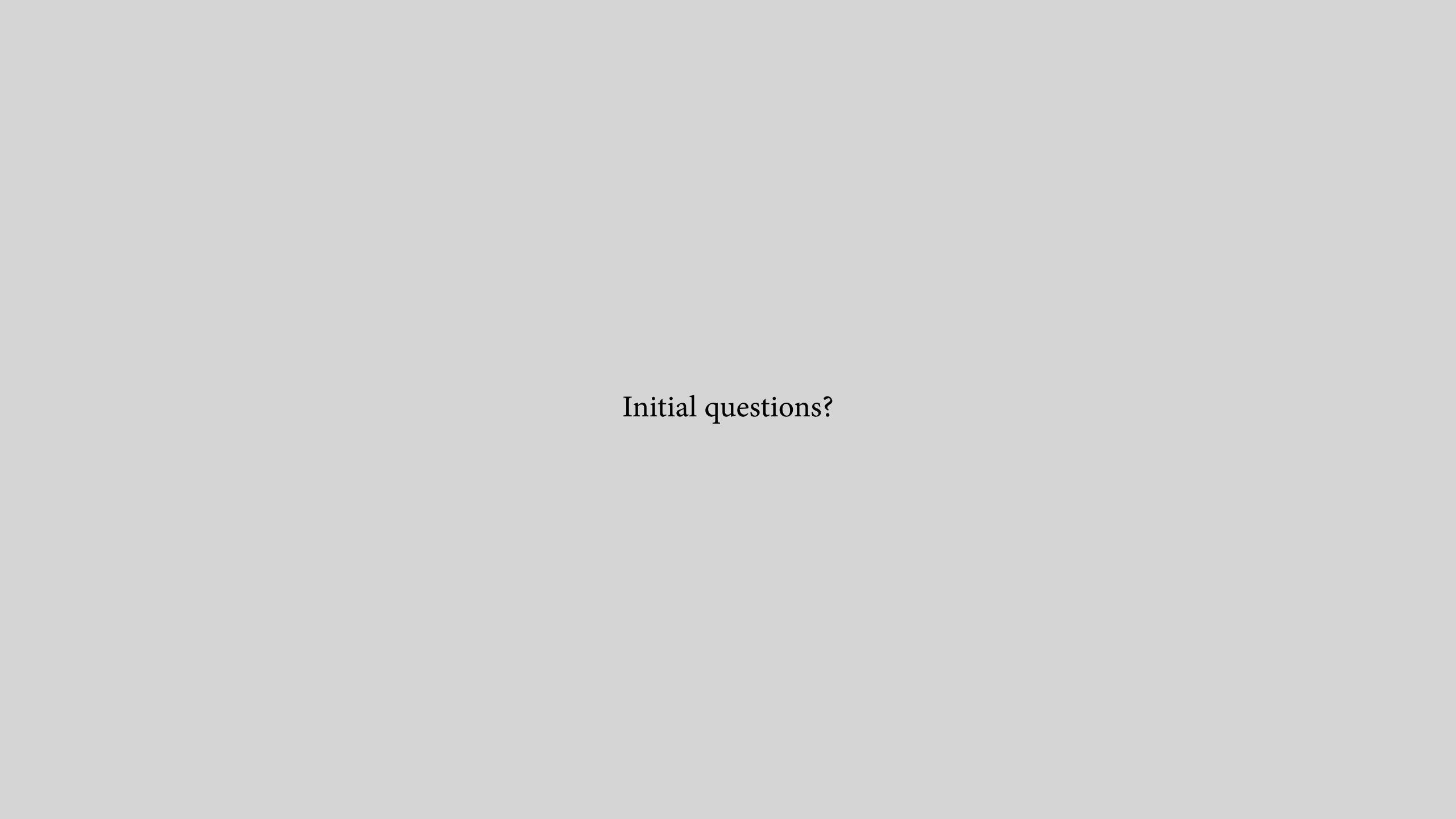
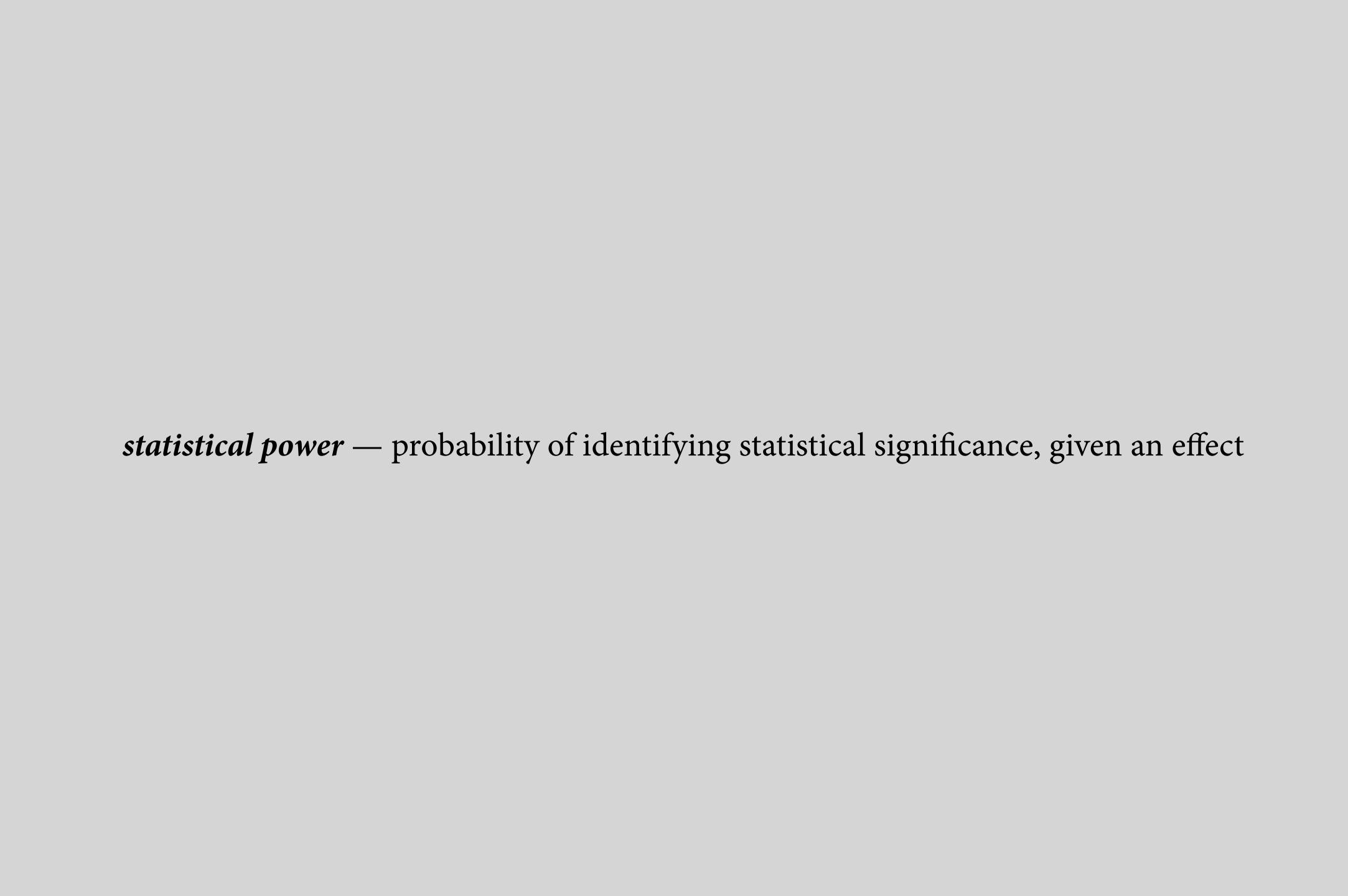
# Research Design

09: statistical power, sample size, simulations

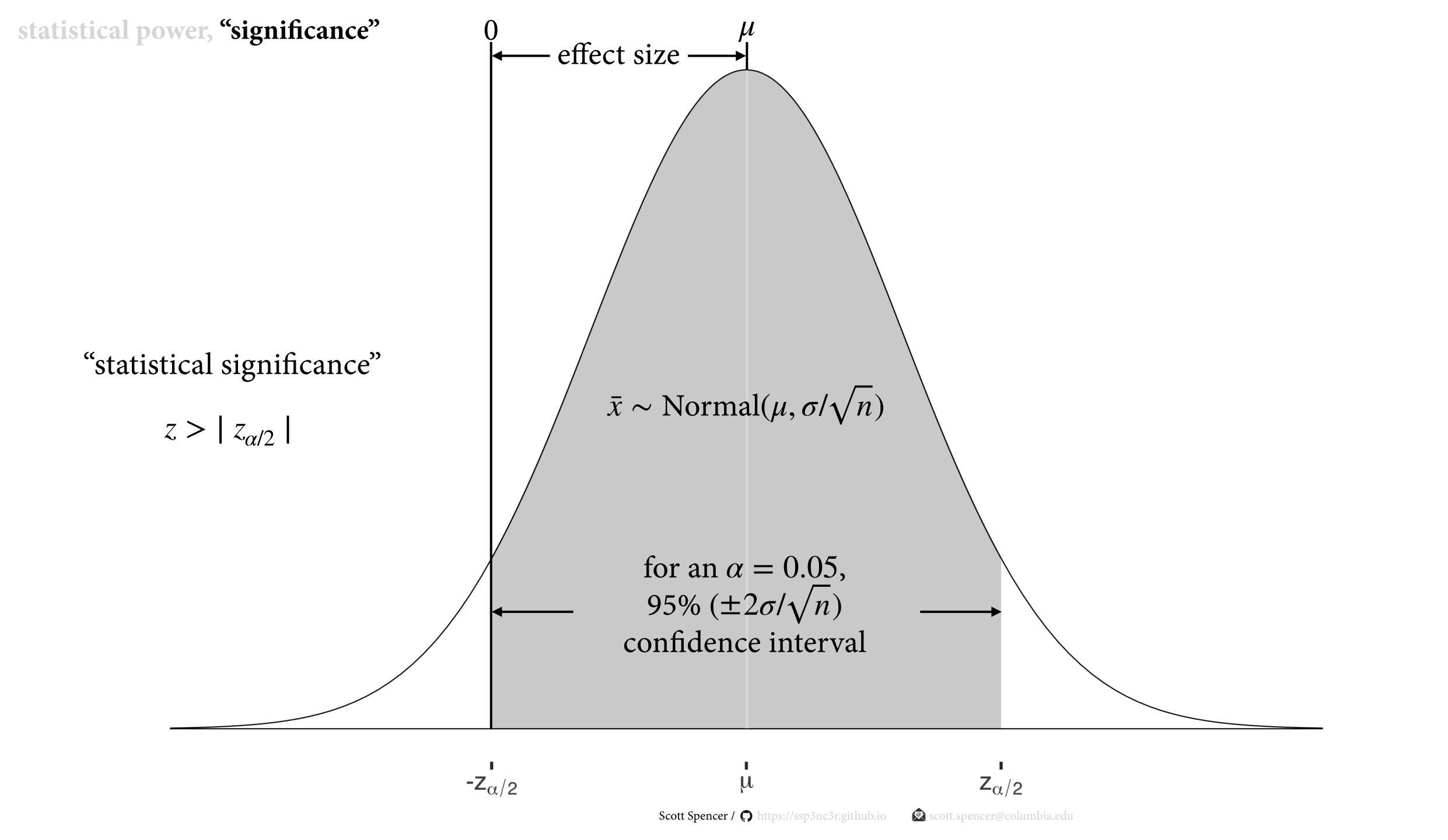


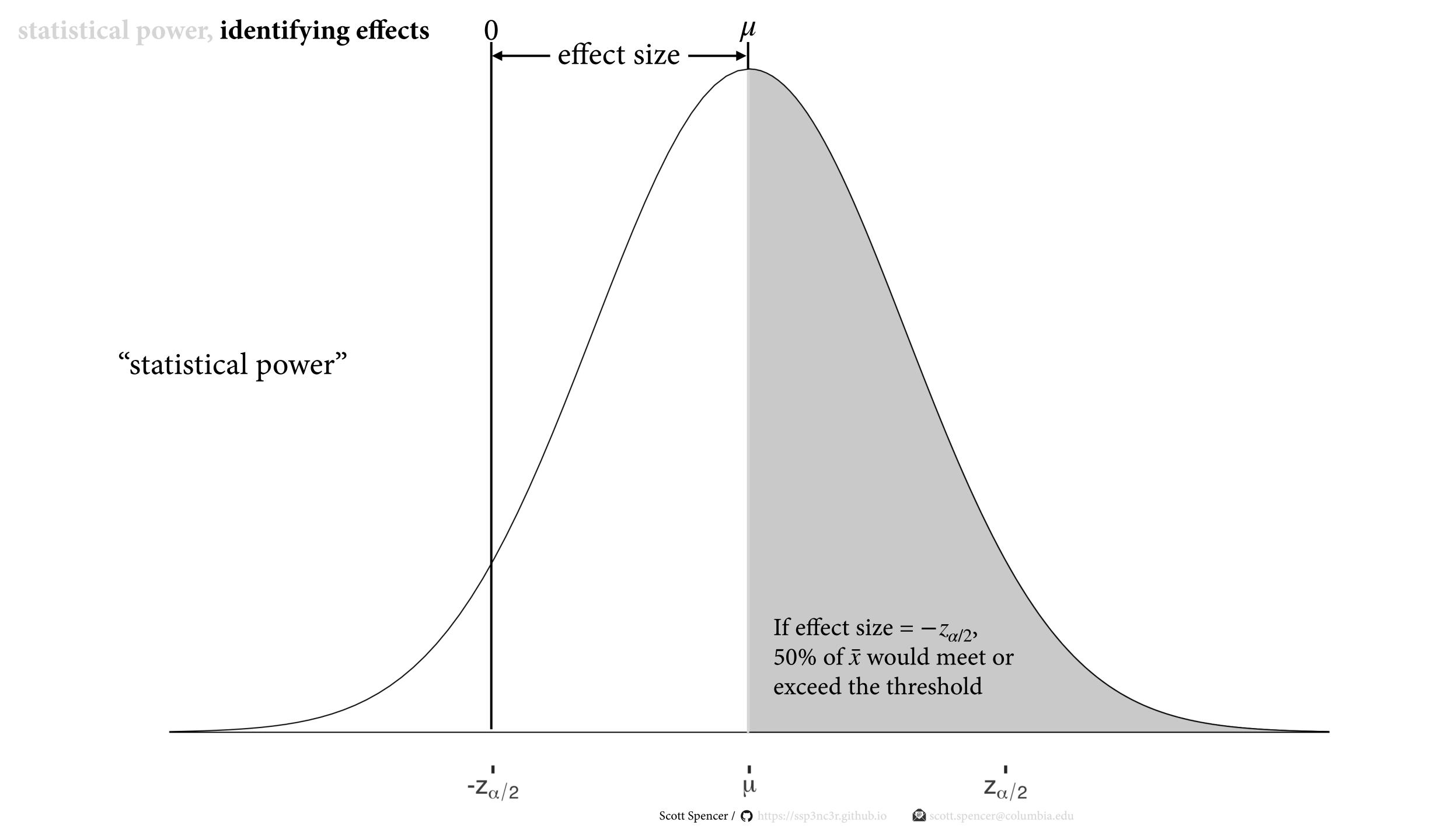


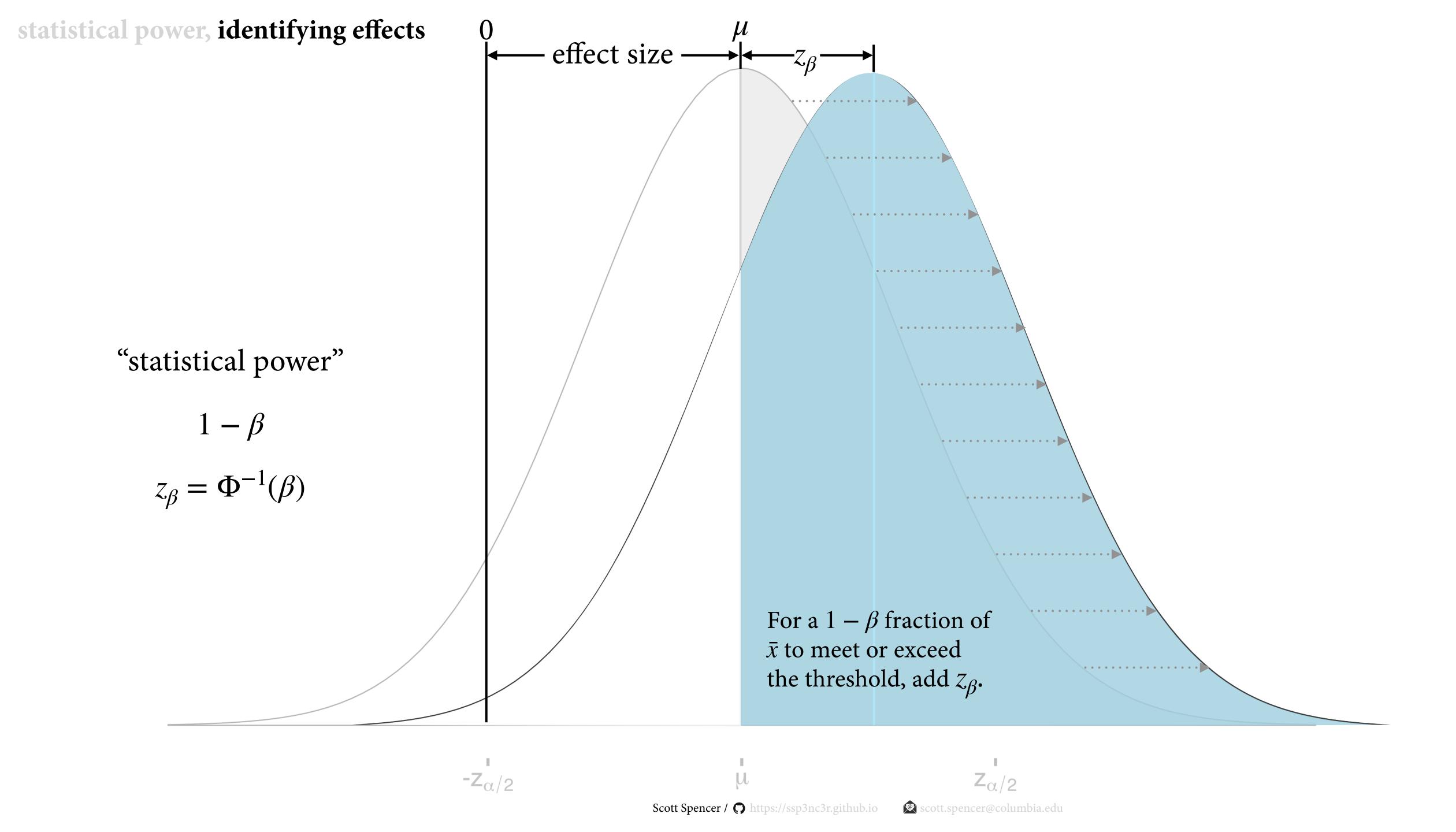
#### statistical power, concept and convention

STATISTICAL POWER | the probability, before a study is performed, that a particular comparison will achieve "statistical significance" at some predetermined level (typically a p-value below 0.05), given some assumed true effect size.

NOTE | a typical threshold for statistical power  $1 - \beta$  is 0.8 but — as with choosing a level of confidence  $\alpha$  — choice  $\beta$  should inform good decisions.







## statistical power, an example — calculating probability of finding an effect

Consider a hypothesis to test

$$H_0: \mu = 0$$
 ,  $H_a: \mu > 0$ 

Choose an appropriate test statistic and reference distribution (probability model)

$$z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}} > |z_{\alpha}| \text{ compared against } F_{\Phi}$$

Choose a meaningful effect size and variation to test

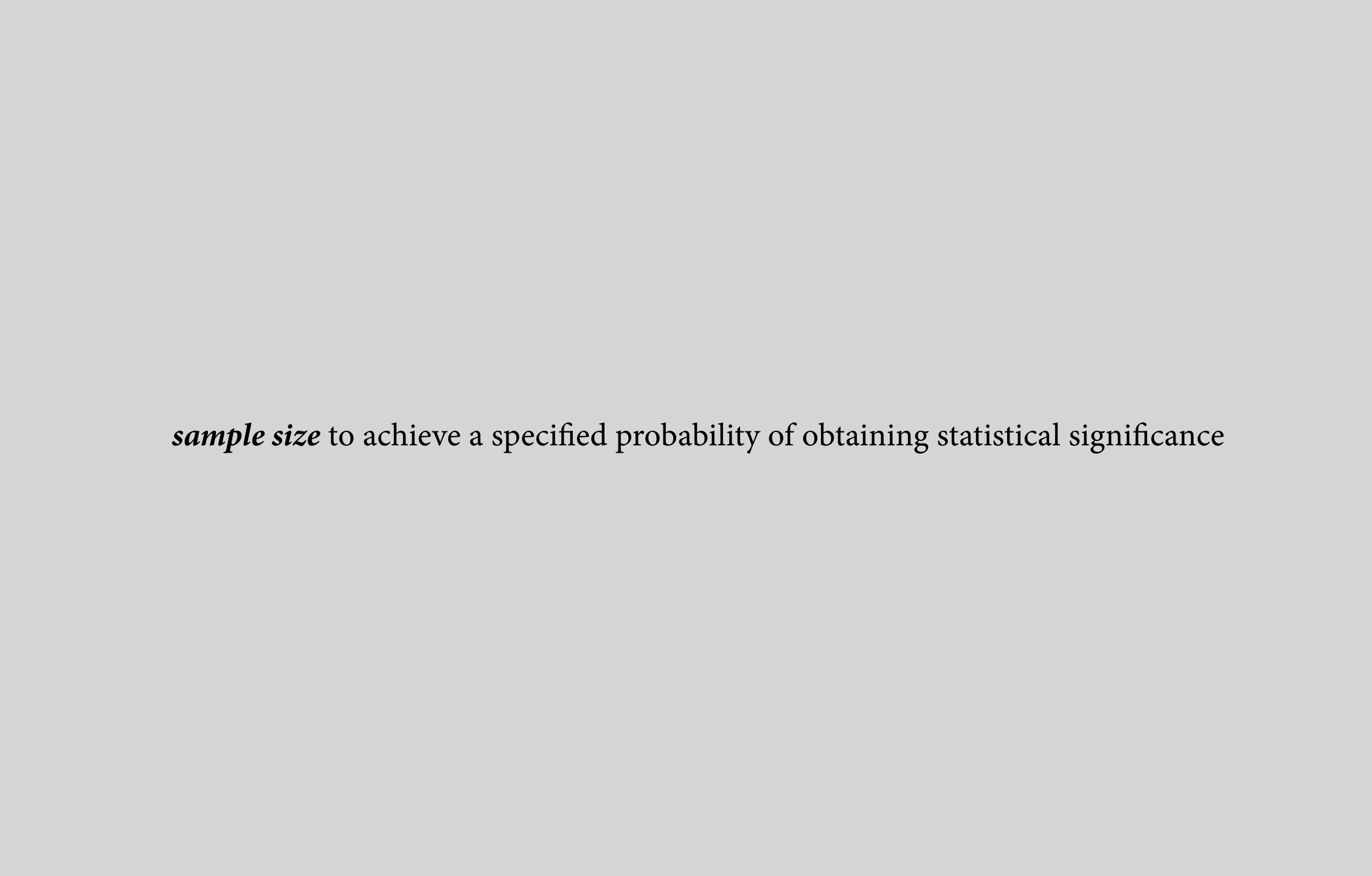
$$\bar{x} > \frac{\sigma}{\sqrt{n}} \mid z_{\alpha} \mid + \mu$$

Choose a sample size

n

Calculate the probability of identifying an underlying effect

$$p(\bar{x} > \frac{\sigma}{\sqrt{n}} \mid z_{\alpha} \mid + \mu) = F_{\Phi}(\frac{\sigma}{\sqrt{n}} \mid z_{\alpha} \mid + \mu)$$



statistical power, estimating sample size to have p chance of finding an effect — solve for n

For a given 
$$\bar{x} - \mu$$
,  $\alpha$ , and  $\beta$ :  $\mu_0 + |Z_{\alpha/2}| \frac{s}{\sqrt{n}} = \bar{x} - |Z_{\beta}| \frac{s}{\sqrt{n}}$ 

$$(|z_{\alpha/2}| + |z_{\beta}|) \cdot \frac{s}{\sqrt{n}} = \bar{x} - \mu_0$$

$$n = \left[ \frac{\left( \left| z_{\alpha/2} \right| + \left| z_{\beta} \right| \right) \cdot s}{\bar{x} - \mu_0} \right]^2$$

### statistical power, estimating sample size — a toy example, estimate sample size for a proportion

For a given 
$$\bar{x} - \mu$$
,  $\alpha$ , and  $\beta$ :  $\mu_0 + |Z_{\alpha/2}| \frac{s}{\sqrt{n}} = \bar{x} - |Z_{\beta}| \frac{s}{\sqrt{n}}$  Let  $\bar{x} = 0.6$ ,  $\mu = 0.5$ ,  $\alpha = 0.05$ , and  $\beta = 0.2$ .

rearrange:

$$(\mid z_{\alpha/2} \mid + \mid z_{\beta} \mid) \cdot \frac{s}{\sqrt{n}} = \bar{x} - \mu$$

$$(|z_{\alpha/2}| + |z_{\beta}|) \cdot \frac{s}{\sqrt{n}} = \bar{x} - \mu_0$$
 
$$(1.96 + 0.84) \frac{\sqrt{0.6(1 - 0.6)}}{\sqrt{n}} = 0.6 - 0.5$$

solve for *n*:

$$n = \left[ \frac{\left( \left| z_{\alpha/2} \right| + \left| z_{\beta} \right| \right) \cdot s}{\bar{x} - \mu_0} \right]^2 \qquad n = \left[ \frac{(1.96 + 0.84) \cdot 0.49}{0.1} \right]^2 = 196$$

$$n = \left[ \frac{(1.96 + 0.84) \cdot 0.49}{0.1} \right]^2 = 196$$



Let  $\bar{x} = 0.6$ ,  $\mu = 0.5$ ,  $\alpha = 0.05$ , and try  $\beta = \{0.2, 0.5\}$ 

Using  $n_{\beta_{0.5}} = 96$  and  $n_{\beta_{0.2}} = 196$ , simulate experiments.

```
p0     <- 0.5
p     <- 0.6
alpha <- 0.05 / 2

z_alpha_2 <- qnorm(alpha, 0, 1, lower.tail = F)

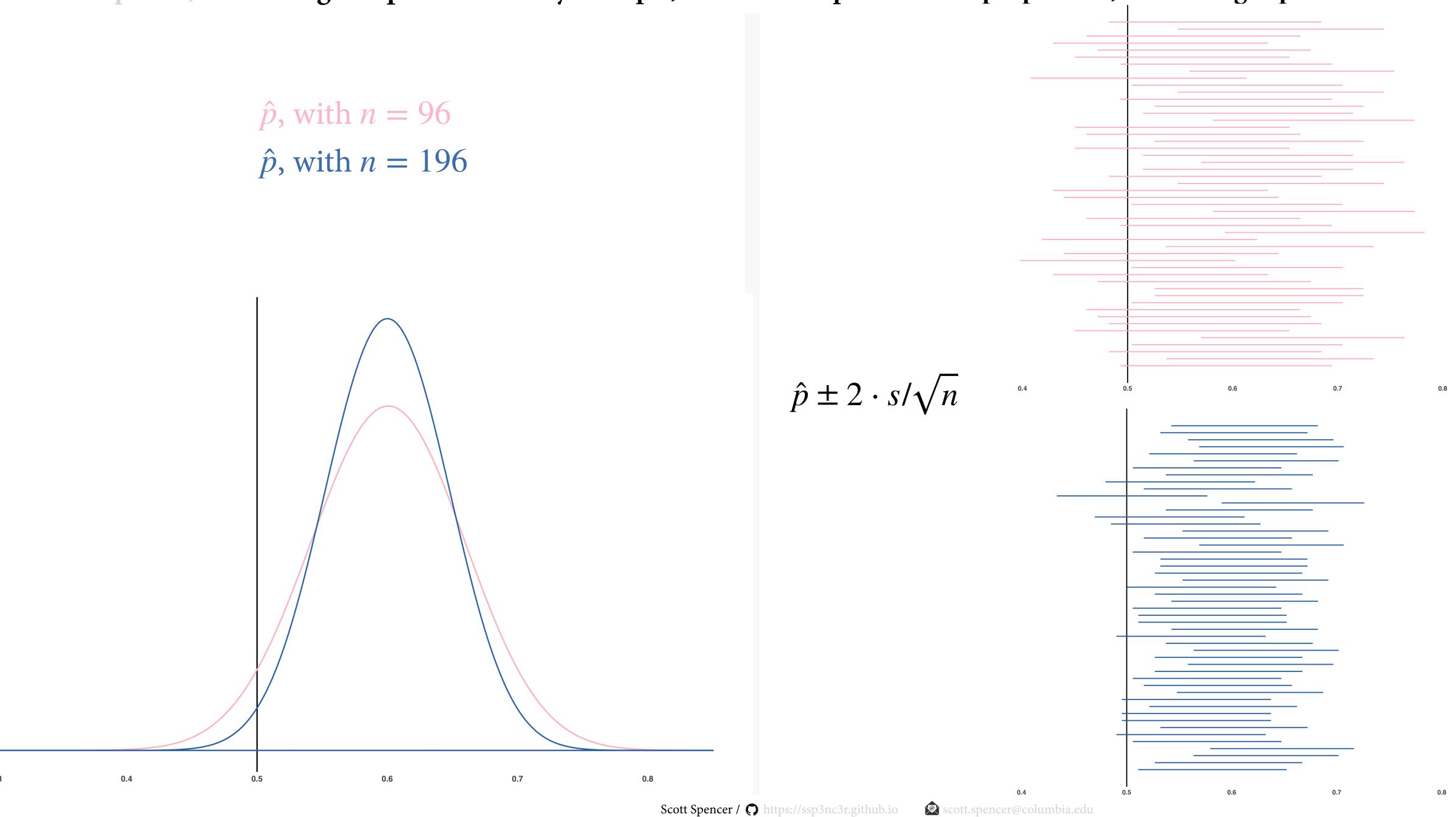
# get n for 80 percent power
beta     <- 0.2
z_beta <- qnorm(1 - beta, 0, 1, lower.tail = T)
n_pwr80 <- ( (z_alpha_2 + z_beta) * sqrt( p0 * (1 - p0) ) / (p - p0) ) ^ 2

# get n for 50 percent power
beta     <- 0.5
z_beta <- qnorm(1 - beta, 0, 1, lower.tail = T)
n_pwr50 <- ( (z_alpha_2 + z_beta) * sqrt( p0 * (1 - p0) ) / (p - p0) ) ^ 2</pre>
```

```
# simulate experiments
survey <- function(n, p) {
    x = ( rbinom(n = n, size = 1, prob = p) )
    x_bar = mean(x)
    se = sd(x) / sqrt(n)
    c(x_bar, se)
}

set.seed(1)
p_hat_96 <- replicate( 1e5, survey(n = n_pwr50, p) )

set.seed(1)
p_hat_196 <- replicate( 1e5, survey(n = n_pwr80, p) )</pre>
```



#### References

Cox, D. R., and N. Reid. "Precision and power, Section 8.1.2." In *The Theory of the Design of* Experiments. Monographs on Statistics and Applied Probability 86. Boca Raton: Chapman & Hall/CRC, 2000.

Gelman, Andrew, and John Carlin. "Beyond Power Calculations." Perspectives on Psychological Science 9, no. 6 (November 2014): 641–51.

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