



# **Introduction to target trial emulation: Worked example**

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# Follow along!

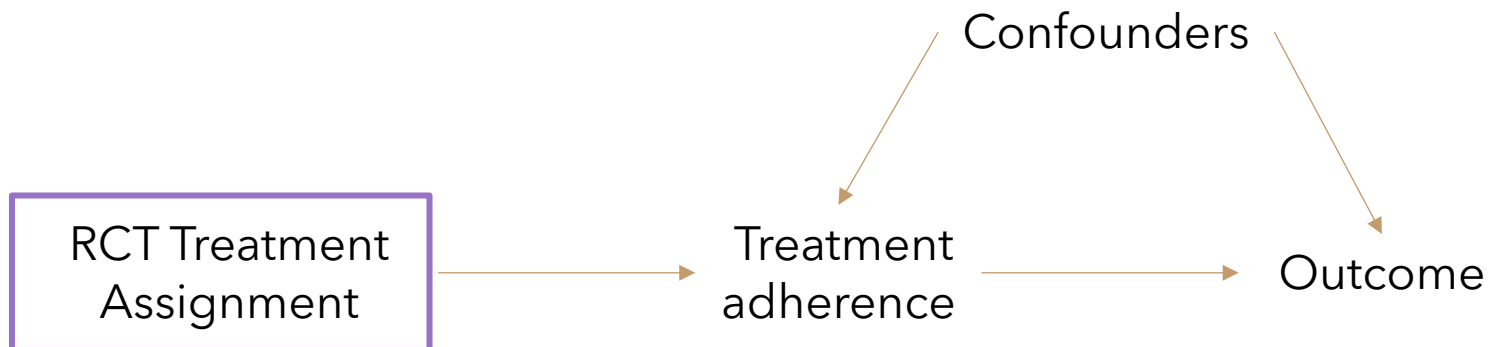
<https://github.com/ssb214/SPER24>

# Study background

- + Caniglia, et al. (2023) – on GitHub
- + **Exposure**: Antibiotic initiation at or after 24 weeks gestation through 36 weeks gestation
- + **Outcome**: Preterm delivery

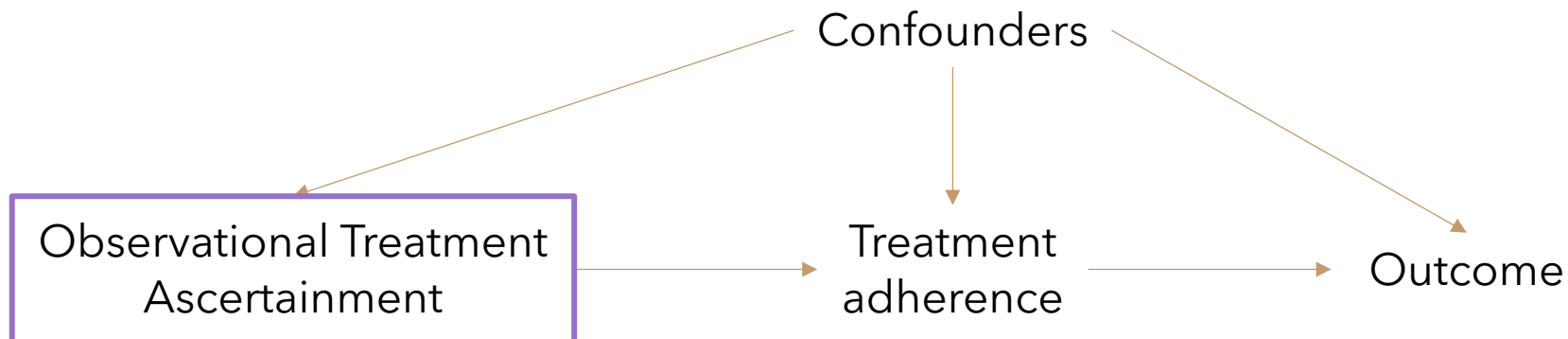
# Causal contrast

- + Intention to treat effect
- + The effect of being assigned to treatment, not the effect of the treatment itself
- + No confounders



# Causal contrast

- + Observational analogue of the **intention to treat** effect
- + Now, we need to control for confounders (in this example, maternal age, parity, and HIV status)
- + The **risk ratio** is the estimate of interest



# “Data” structure

- + Define eligibility criteria
- + Specify time zero
- + Determine treatment ascertainment
- + Confounders
- + Follow-up and outcome

participant_id	age	parity	hiv	week	exposure_info	outcome
1	18-35	1	0	wk_24	0	0
1	18-35	1	0	wk_25	0	1
1	18-35	1	0	wk_26	0	NA
1	18-35	1	0	wk_27	0	NA
1	18-35	1	0	wk_28	0	NA
1	18-35	1	0	wk_29	0	NA
1	18-35	1	0	wk_30	0	NA
1	18-35	1	0	wk_31	0	NA
1	18-35	1	0	wk_32	0	NA
1	18-35	1	0	wk_33	0	NA
1	18-35	1	0	wk_34	0	NA
1	18-35	1	0	wk_35	0	NA
1	18-35	1	0	wk_36	0	NA
2	18-35	0	0	wk_24	0	0
2	18-35	0	0	wk_25	0	0
2	18-35	0	0	wk_26	0	0
2	18-35	0	0	wk_27	0	0
2	18-35	0	0	wk_28	0	0



# Code overview



<https://github.com/ssb214/SPER24>

**Step 1: estimate a RR for each target trial within week 24-36**



```
adj_model_results[[wk]] <- glm(outcome ~ exposure_info +  
  hiv +  
  parity +  
  age,  
  data = data_wk,  
  family = binomial(link = "log")) # log link for RR  
}
```

**General model for weeks 24-36**

# Quick note: Families and links

Odds Ratio	Risk Ratio	Risk Difference
Family = Binomial	Family = Binomial	Family = Binomial
Link = Logistic	Link = Log	Link = Identity
SE = model-based	SE = model-based	SE = model-based
	Family = Poisson	Family = Gaussian
	Link = Log	Link = Identity
	SE = Bootstrap	SE = Bootstrap

# Sequential target trials within each week

```
# Create a list to store results for each week
adj_model_results <- list()

for (wk in paste0("wk_", 24:36)) {
  # Subset data for the current week
  data_wk <- final_data %>% filter(week == wk)

  # Fit log-binomial regression model and store results
  adj_model_results[[wk]] <- glm(outcome ~ exposure_info +
                                hiv +
                                parity +
                                age,
                                data = data_wk,
                                family = binomial(link = "log")) # log link for RR
}
```

# Pulling results

```
# Pull RR (exp. beta values) and 95% confidence intervals into a dataframe
## Create a null dataframe
results_adj <- data.frame(
  Week = character(),
  Coefficient = numeric(),
  Lower_CI = numeric(),
  Upper_CI = numeric(),
  stringsAsFactors = FALSE
)

for (wk in paste0("wk_", 24:36)) {
  # Extract coefficients and confidence intervals
  coefficients <- exp(crude_model_results[[wk]]$coefficients)
  ci <- exp(confint(crude_model_results[[wk]]))

  # Store results in the data frame
  results_adj <- rbind(results_adj, data.frame(
    Week = wk,
    Coefficient = coefficients["exposure_info"],
    Lower_CI = ci["exposure_info", "2.5 %"],
    Upper_CI = ci["exposure_info", "97.5 %"]
  ))
}
```

**Step 2: estimate RR for pooled results across weeks 24-36**


# Pooled model

```
library(splines)

model_pooled <- glm(outcome ~ exposure_info +
                    hiv +
                    parity +
                    age +
                    ns(week_num, df = 2), # adding flexibility to the model for GA
                    data = final_data,
                    family = binomial(link = "log")) # log link for RR
```

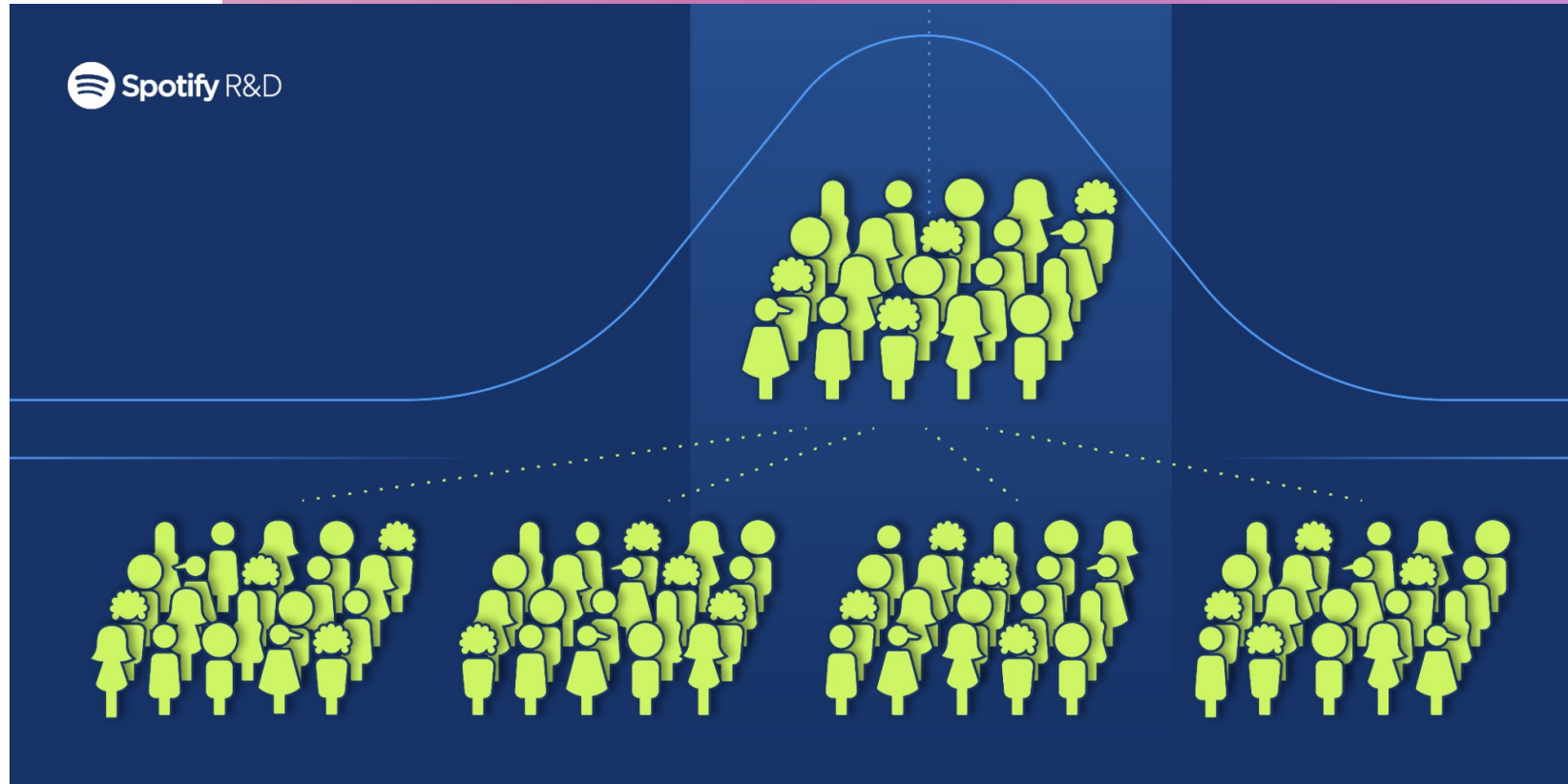
# Pooled results

- + Now we have the Beta estimates for the model, but because people can be present in more than one trial, we need to use the **bootstrap** for SE



Coefficients:				
	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-1.969904	0.017202	114.515	<2e-16 ***
exposure_info	0.023687	0.027931	0.848	0.396
hiv1	-0.009860	0.007708	-1.279	0.201
parity1	0.008752	0.006526	1.341	0.180
age>35	0.001125	0.017756	0.063	0.949
age18-35	0.001545	0.016367	0.094	0.925
ns(week_num, df = 2)1	-0.449437	0.015827	-28.415	<2e-16 ***
ns(week_num, df = 2)2	-0.696855	0.017062	-40.842	<2e-16 ***

# The bootstrap





```
replicates <- 25 # this needs to be increased for a real analysis  
bootstrap_estimates <- NULL
```

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replicates <- 25 # this needs to be increased for a real analysis
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for (i in 1:replicates) {
  # set seed to i so we get a unique sample each time
  set.seed(i)

  # sample the data
  index <- sample(1:nrow(final_data), nrow(final_data), replace = T)

  final_data_sample <- final_data[index,]
```

```
> index[1:10]
[1] 452737 124413 856018 25173 1343338 1441262 640775 538191 1318949 549593
```

# The bootstrap

```
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for (i in 1:replicates) {
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  set.seed(i)

  # sample the data
  index <- sample(1:nrow(final_data), nrow(final_data), replace = T)

  final_data_sample <- final_data[index,]

  # estimate the association in the resample
  model_boot <- glm(outcome ~ exposure_info +
                    hiv +
                    parity +
                    age +
                    ns(week_num, df = 2), # adding flexibility to the model for week
                    data = final_data_sample,
                    family = binomial(link = "log")) # log link for RR

  bootstrap_estimates <- rbind(
    bootstrap_estimates,
    coef(model_boot)[2]
  )

  # Print iteration number when i is divisible by 10
  if (i %% 10 == 0) {
    print(paste("Iteration:", i))
  }
}
```

# Pooled results

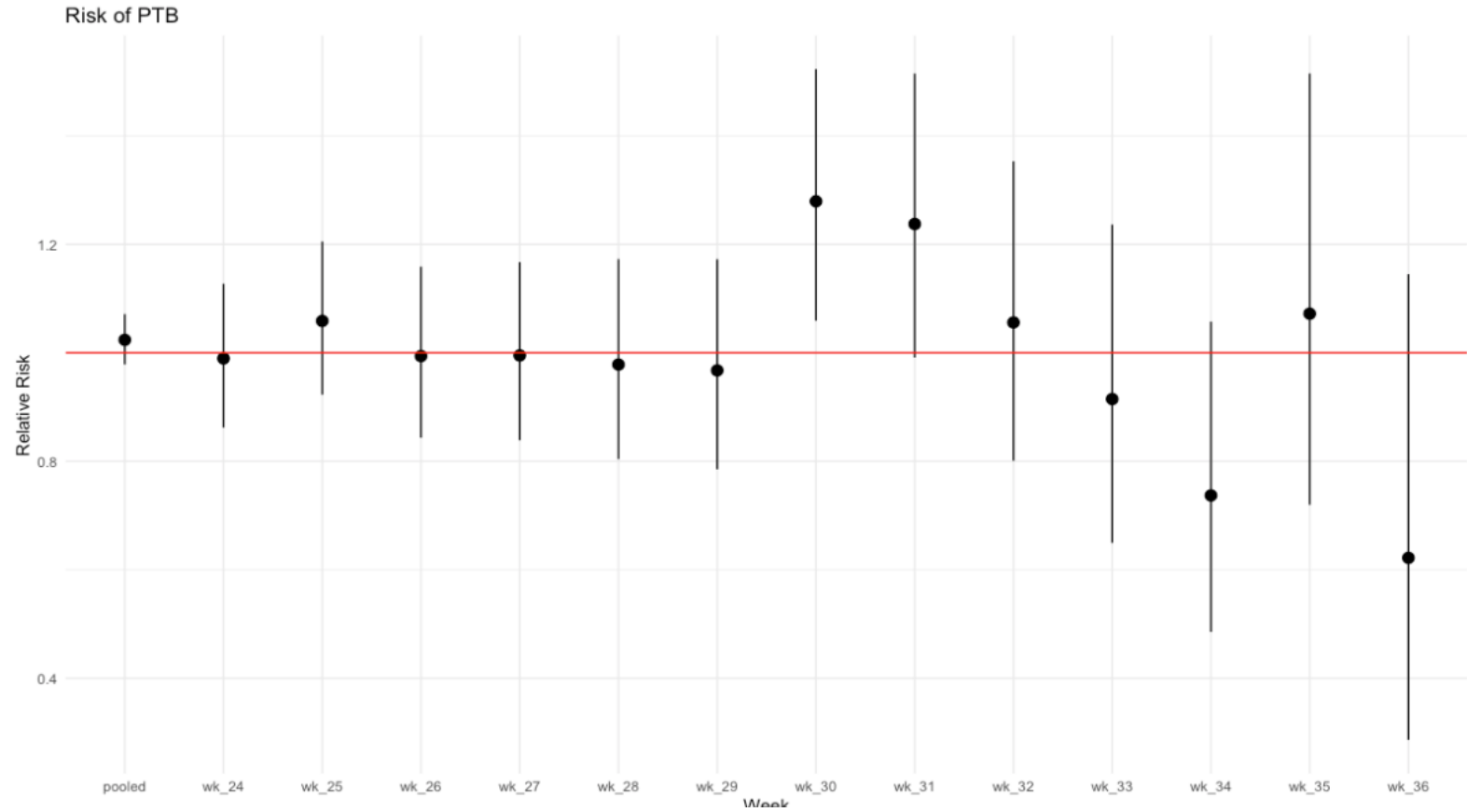
```
se_estimate <- sd(bootstrap_estimates)

lcl_pooled <- exp(coef(model_pooled)[2] - 1.96 * se_estimate)
ucl_pooled <- exp(coef(model_pooled)[2] + 1.96 * se_estimate)

pooled_dat <- data.frame(
  Week = "pooled",
  Coefficient = co_pooled[2],
  Lower_CI = lcl_pooled,
  Upper_CI = ucl_pooled
)

results_adj <- rbind(results_adj, pooled_dat)
```

Antibiotic Initiation at:	Initiation	# of individuals	Events, #	RR (95% CI)
24 weeks	No	110082	15659	0.99
	Yes	1321	186	(0.86, 1.12)
25 weeks	No	108628	13186	1.06
	Yes	1454	186	(0.92, 1.21)
26 weeks	No	107316	10975	0.99
	Yes	1312	133	(0.84, 1.16)
27 weeks	No	105876	8963	0.99
	Yes	1440	124	(0.84, 1.17)
28 weeks	No	104555	7679	0.98
	Yes	1321	95	(0.80, 1.17)
29 weeks	No	103095	6289	0.96
	Yes	1460	85	(0.79, 1.17)
30 weeks	No	101656	5265	1.28
	Yes	1439	100	(1.06, 1.52)
31 weeks	No	100249	4446	1.24
	Yes	1407	74	(0.99, 1.51)
32 weeks	No	98962	3636	1.06
	Yes	1287	50	(0.80, 1.35)
33 weeks	No	97722	2972	0.91
	Yes	1240	34	(0.65, 1.24)
34 weeks	No	1317	2327	0.74
	Yes	96405	24	(0.49, 1.06)
35 weeks	No	1303	1691	1.07
	Yes	95102	26	(0.72, 1.52)
36 weeks	No	93841	1037	0.62
	Yes	1261	8	(0.29, 1.15)
Pooled (24-36 weeks)	No	1323489	84125	1.02
	Yes	17562	1125	(0.98, 1.07)



# Questions?

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