

# Your Paper Title Here

Your Name(s)

Date

## **Abstract**

Write your abstract here.

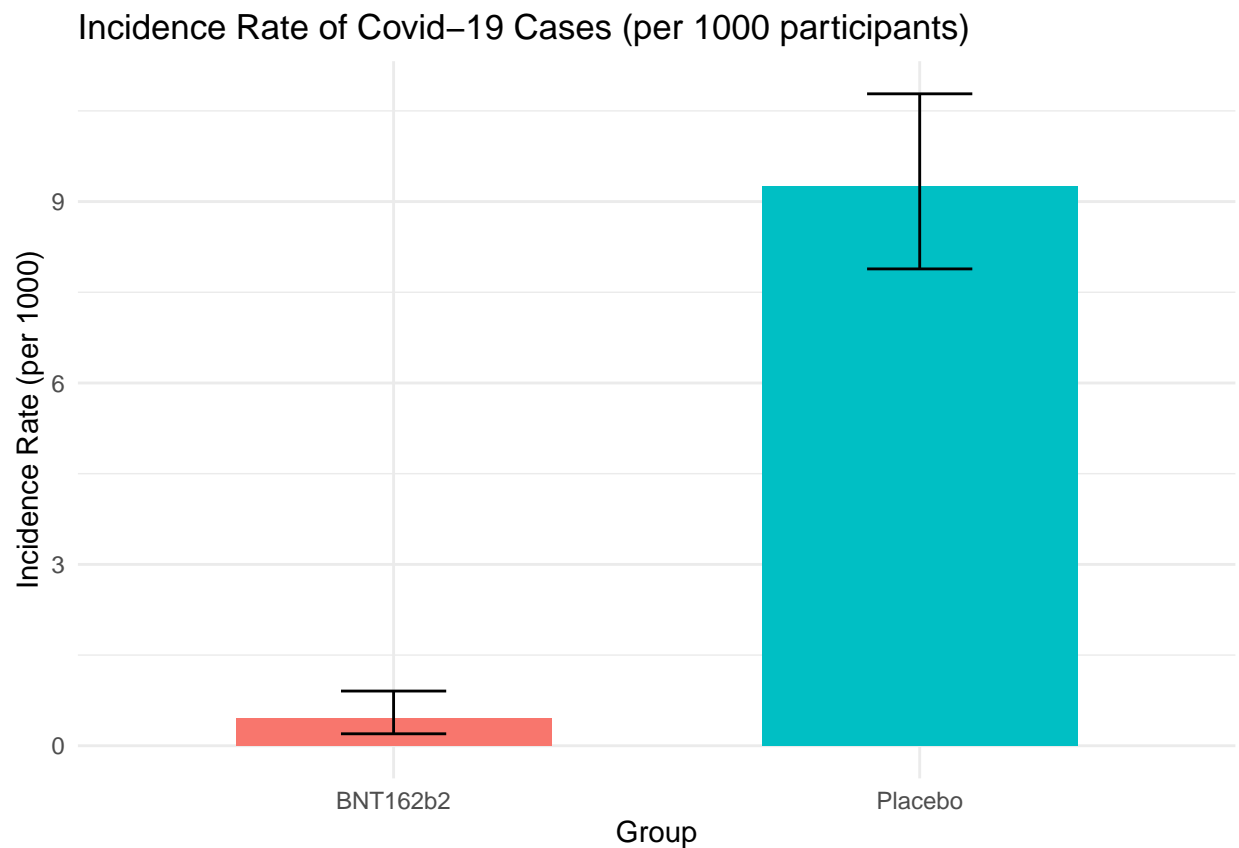
***TESTTTING\****

## **Keywords**

*Keyword 1, Keyword 2, Keyword 3, Keyword 4*

## Introduction / Background

Provide an introduction with background information.



## Estimated Vaccine Efficacy (VE): 95.03 %

## Statistical Methods

### Model

Describe the statistical model used.

### Likelihood Inference

Detail the likelihood approach.

#likelihood function:

#MLE

#Confidence Interval for Psi

#Large sample confidence interval

#Bootstrap

We will also use a computational approach to find our  $\hat{\psi}_{MLE}$  interval by bootstrapping our  $\hat{\pi}$ . We will randomly draw with replacement many times from a  $Binomial(n, \pi_{observed})$  distribution. We will then calculate a value of  $\hat{\psi}_i$  for each  $\hat{\pi}_i$  from our bootstrap. We will then construct a 95% confidence interval by calculating the 2.5th and 97.5th percentile of our  $\hat{\psi}$ s

## Bayesian Inference

Detail the Bayesian approach.

## Results

Present your findings.

Our bootstrap 95% confidence interval is:

##        2.5%        97.5%

## 0.9115646 0.9808917

We will use a beta binomial model:

$$T_{\pi} \sim Binom(170, \pi) \quad g(\pi) = Beta(a, b)$$

$$P(\psi \geq 0.3) = P\left(\frac{1-2\pi}{1-\pi} \geq 0.3\right) = P\left(\pi \leq \frac{7}{17}\right) = \frac{7}{17}$$

for our model we will use the following priors

$$\text{experimental prior } P(\psi \geq 0.3) \approx 0.57 \quad P(\psi \geq 0) = 1$$

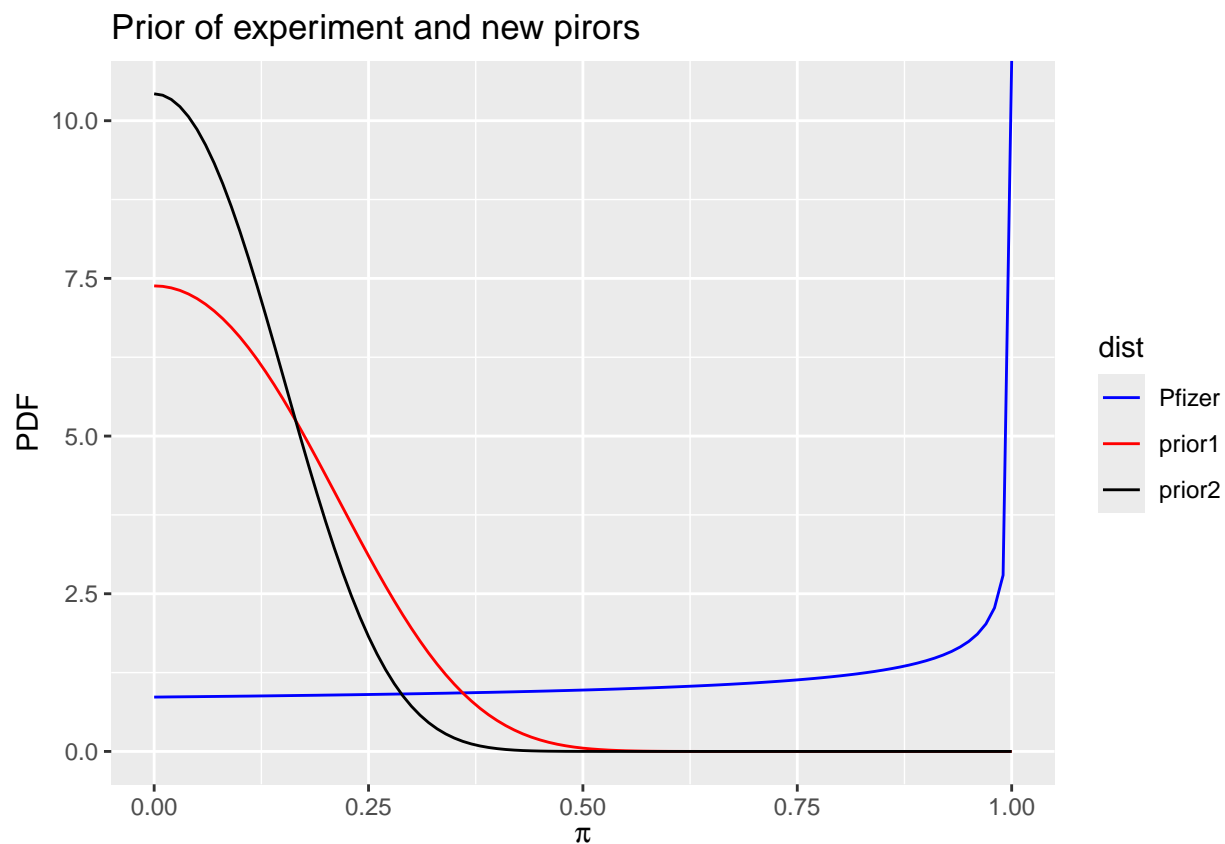
$$0.05 \text{ prior } P(\psi \geq 0.3) = 0.05 \quad P(\psi \geq 0) = 0.5$$

$$0.01 \text{ prior } P(\psi \geq 0.3) = 0.01 \quad P(\psi \geq 0) = 0.5$$

The experimental prior is the prior that Pfizer used in their study, which gives the vaccine the best chance of passing the tests. The following two priors have a decreasing chance of passing the tests.

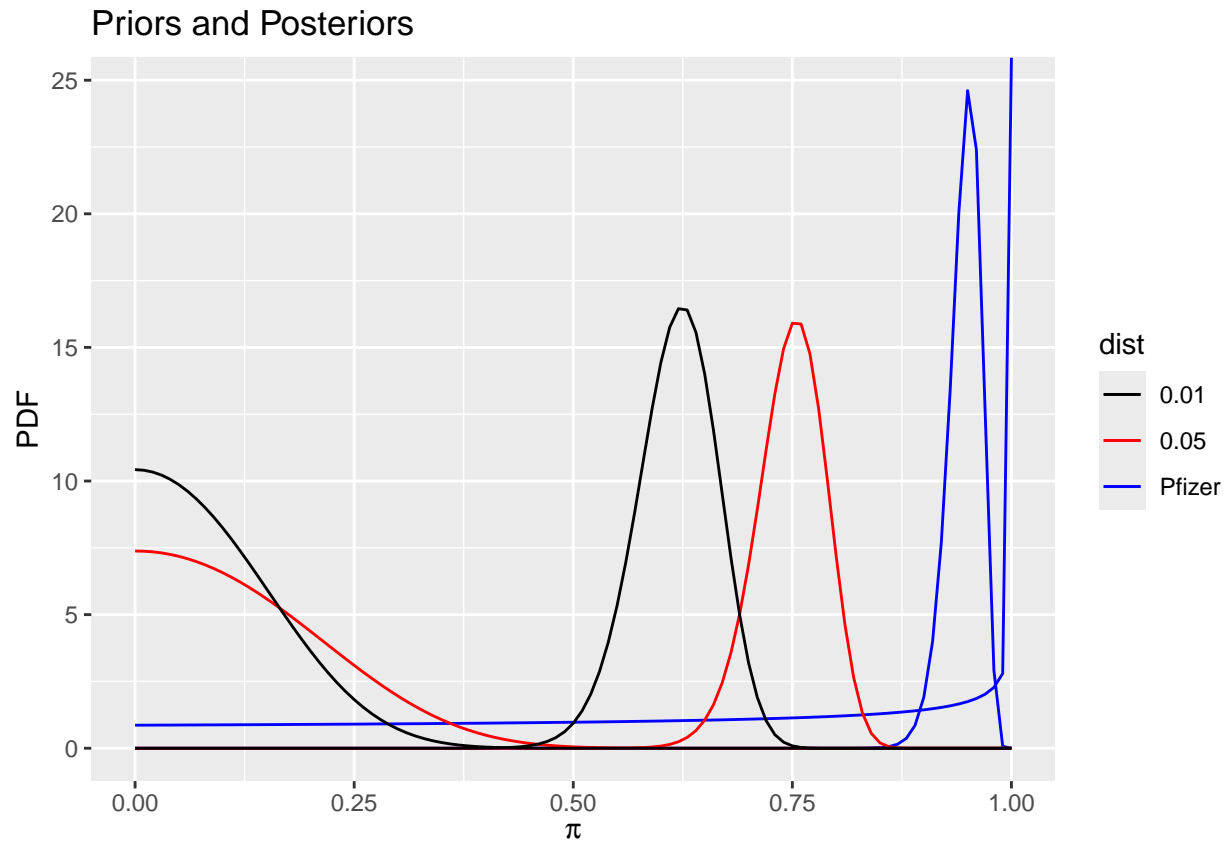
The Pfizer study used  $Beta(0.700102, 1)$  as their prior Betas. We will be trying  $Beta(43.03, 43.03)$  and  $Beta(85.63, 85.63)$  as our prior Beta values.

We can visualize how much more likely their prior is than ours



So, the posterior distributions are as follows: Pfizer:  $h(\pi|t) = Beta( 8.700102, 163)$  0.05 posterior:  $h(\pi|t) = Beta( 51.03, 205.03)$  0.01 posterior:  $h(\pi|t) = Beta( 93.63, 247.63)$

We can visualize the comparison between all of the priors and posteriors:



summarizing the posterior distribution of  $\pi$

```
## Pfizer posterior median: 0.0489
```

```
## 95% CI for Pfizer: [ 0.0232 , 0.088 ]
```

```
## 0.05 prior posterior median: 0.1985 0.1985
```

```
## 95% CI for 0.05 prior: [ 0.1528 , 0.2503 ]
```

```
## 0.01 prior posterior median: 0.2739 0.2739
```

```
## 95% CI for 0.01 prior: [ 0.2284 , 0.3228 ]
```

## Significance Tests and Posterior Distributions

The Pfizer study used a  $\text{Beta}(0.700102, 1)$  prior. In our analysis, we compare this with two alternative prior distributions:

$$\text{Beta}\left(\text{Prior 1: Beta}(43.03, 43.03)\right) \quad \text{and} \quad \text{Beta}\left(\text{Prior 2: Beta}(85.63, 85.63)\right).$$

Given the data (with 8 successes and 162 failures), the posterior distributions are as follows:

- **Pfizer Posterior:**

$$h(\pi \mid t) = \text{Beta}(0.700102 + 8, 162 + 1)$$

- **0.05 Prior Posterior:**

$$h(\pi \mid t) = \text{Beta}(43.03 + 8, 162 + 43.03)$$

- **0.01 Prior Posterior:**

$$h(\pi \mid t) = \text{Beta}(85.63 + 8, 162 + 85.63)$$

Below we compute the posterior medians and approximate 95% credible intervals (using the 2.5th and 97.5th percentiles):

- **Pfizer Posterior:**

Posterior median:  $\approx 0.0489$

95% CI:  $[0.0232, 0.088]$

- **0.05 Prior Posterior:**

Posterior median:  $\approx 0.1985$

95% CI:  $[0.1528, 0.2503]$

- **0.01 Prior Posterior:**

Posterior median:  $\approx 0.2739$

95% CI:  $[0.2284, 0.3228]$

## Hypothesis Tests

We test the hypothesis for vaccine effectiveness:

$$H_0 : \pi \geq 0.5 \quad \text{vs} \quad H_1 : \pi < 0.5.$$

For each posterior distribution, the p-value is computed as the probability of observing  $\pi \geq 0.5$  under the posterior, i.e.,

$$p\text{-value} = 1 - F(0.5),$$

where  $F(\beta)$  is the cumulative distribution function of the corresponding Beta posterior.

Pfizer Posterior:

$$p\text{-value} \approx 0.$$

0.05 Prior Posterior:

$$p\text{-value} \approx 0.$$

0.01 Prior Posterior:

$$p\text{-value} \approx 0.$$

## Discussion / Conclusion

Discuss / conclude here.

## Bibliography

Brown, B. (2024). *Lecture Title*. Lecture slides, Course Name, University Name.

Doe, J. (2020). Title of the Paper. *Journal Name*, 12(3), 45-67.

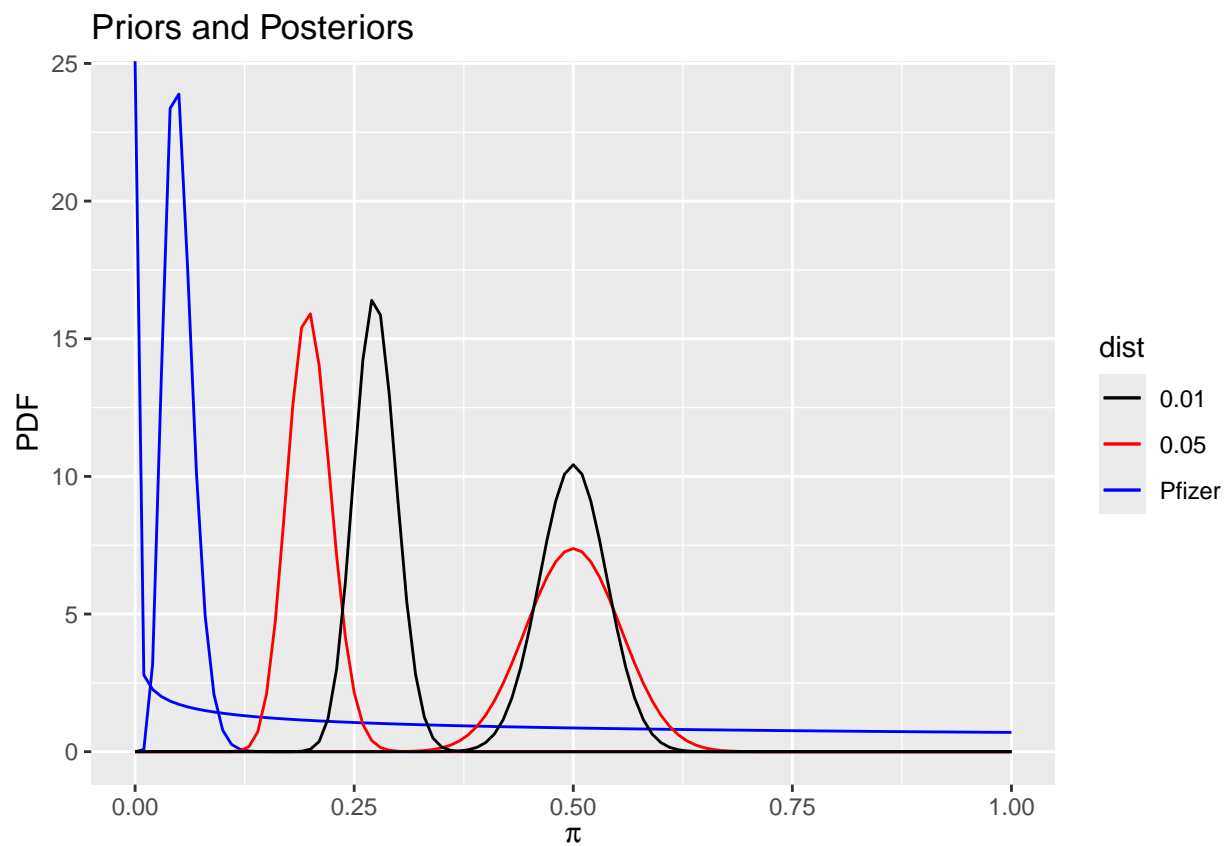
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## Appendix

### Code

### Plots





## Proofs

If applicable, include detailed mathematical derivations or additional theoretical explanations.

Proof of the Maximum Likelihood Estimator for  $\hat{\psi}_0^{mle}$ :

We can begin with the Likelihood function. Recall that the design of our experiment is binomial; we observe 170 infections and try to make inference on  $\pi$ , the proportion of infections from those who have the vaccine. Assuming each observation is independent (which is necessary for a binomial model), we can write our likelihood function in terms of  $\pi$ :

$$L(\pi) = \binom{170}{x} (\pi)^x (1 - \pi)^{170-x}$$

Where  $x$  is the observed number of infections where the patient had the vaccine. We observe  $x = 8$ , so we can rewrite our likelihood function:

$$L(\pi; x = 8) = \binom{170}{8} (\pi)^8 (1 - \pi)^{162}$$

Notably, however, we are not hoping to draw inference on  $\pi$ , but rather on  $\psi$ , the efficacy of the vaccine. We can write this parameter in terms of  $\pi$ .

$$\psi = \frac{1 - 2\pi}{1 - \pi}$$

In order to write our Likelihood function in terms of  $\psi$ , we can rewrite  $\psi$  in terms of  $\pi$ :

$$\pi = \frac{\psi - 1}{\psi - 2}$$

Now we rewrite our likelihood function.

$$L^*(\psi) = \binom{170}{8} \left( \frac{\psi - 1}{\psi - 2} \right)^8 \left( \frac{-1}{\psi - 2} \right)^{162}$$

Now that we have written our full likelihood function in terms of  $\psi$ , we can write the log likelihood

function.

$$\ell^*(\psi) = \log \binom{170}{8} + 8 \log(\psi - 1) - 8 \log(\psi - 2) - 162 \log(\psi - 2) + 162 \log(-1)$$

We have  $162 \log(-1) = \log((-1)^{162}) = \log(1) = 0$ :

$$\ell^*(\psi) = \log \binom{170}{8} + 8 \log(\psi - 1) - 170 \log(\psi - 2)$$

To maximize this equation, we take its derivative with respect to  $\psi$ :

$$\frac{d}{d\psi} \ell^*(\psi) = \frac{8}{\psi - 1} - \frac{170}{\psi - 2}$$

Setting this derivative finds the maximum:

$$0 = \frac{8}{\hat{\psi}_0^{mle} - 1} - \frac{170}{\hat{\psi}_0^{mle} - 2}$$

$$\frac{170}{\hat{\psi}_0^{mle} - 2} = \frac{8}{\hat{\psi}_0^{mle} - 1}$$

$$170\hat{\psi}_0^{mle} - 170 = 8\hat{\psi}_0^{mle} - 16$$

$$\hat{\psi}_0^{mle} = \frac{154}{162}$$