

# 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.

## Statistical Methods

### Model

Describe the statistical model used.

### Likelihood Inference

Detail the likelihood approach.

#likelihood function:

#MLE

#Confidence Interval for Phi

#Large sample confidence interval

#Bootstrap

```
## [1] 6 10 7 11 12 3 8 11 8 7 12 7 9 8 4 11 6 3 6 12 11 9 8 15 8
## [26] 9 8 8 6 5 13 11 9 10 3 7 9 5 6 5 5 7 7 6 5 5 6 7 6 10
## [51] 3 7 10 4 8 5 5 9 11 7 9 4 7 6 10 7 10 10 10 7 9 8 9 1 7
## [76] 5 7 8 6 4 6 9 7 10 4 7 14 11 11 5 5 8 6 8 6 5 10 4 7 7
```

```
## [1] 0.9503042
```

```
## [1] 0.9610390 0.9333333 0.9542484 0.9261745 0.9189189 0.9808917 0.9473684
## [8] 0.9261745 0.9473684 0.9542484 0.9189189 0.9542484 0.9403974 0.9473684
## [15] 0.9743590 0.9261745 0.9610390 0.9808917 0.9610390 0.9189189 0.9261745
## [22] 0.9403974 0.9473684 0.8965517 0.9473684 0.9403974 0.9473684 0.9473684
```

```
## [29] 0.9610390 0.9677419 0.9115646 0.9261745 0.9403974 0.9333333 0.9808917
## [36] 0.9542484 0.9403974 0.9677419 0.9610390 0.9677419 0.9677419 0.9542484
## [43] 0.9542484 0.9610390 0.9677419 0.9677419 0.9610390 0.9542484 0.9610390
## [50] 0.9333333 0.9808917 0.9542484 0.9333333 0.9743590 0.9473684 0.9677419
## [57] 0.9677419 0.9403974 0.9261745 0.9542484 0.9403974 0.9743590 0.9542484
## [64] 0.9610390 0.9333333 0.9542484 0.9333333 0.9333333 0.9333333 0.9542484
## [71] 0.9403974 0.9473684 0.9403974 0.9937107 0.9542484 0.9677419 0.9542484
## [78] 0.9473684 0.9610390 0.9743590 0.9610390 0.9403974 0.9542484 0.9333333
## [85] 0.9743590 0.9542484 0.9041096 0.9261745 0.9261745 0.9677419 0.9677419
## [92] 0.9473684 0.9610390 0.9473684 0.9610390 0.9677419 0.9333333 0.9743590
## [99] 0.9542484 0.9542484
```

```
##      2.5%      97.5%
```

```
## 0.9150579 0.9808917
```

## Bayesian Inference

Detail the Bayesian approach.

```
#beta binomial model  $T_\pi \sim \text{Binom}(170, \pi)$   $g(\pi) = \text{Beta}(a, b)$ 
```

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

```
#priors We will use the following priors
```

experimental prior  $P(\psi \geq 0.3) = 0.700102$   $P(\psi \geq 0) = 1$

0.05 prior  $P(\psi \geq 0.3) = 0.05$   $P(\psi \geq 0) = 0.5$

0.01 prior  $P(\psi \geq 0.3) = 0.01$   $P(\psi \geq 0) = 0.5$

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.

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

```
#betas
```

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. So, the posterior distributions are as follows:  
Pfizer:  $h(\pi|t) = Beta(8.700102, 162.700102)$  0.05 posterior:  $h(\pi|t) = Beta(51.03, 205.03)$  0.01  
posterior:  $h(\pi|t) = Beta(93.63, 247.63)$

summarizing the posterior distribution of  $\pi$

```
## posterior median 0.3877
```

```
## [1] 0.3394897 0.4374038
```

```
## Warning: package 'HDInterval' was built under R version 4.4.3
```

```
##      lower      upper
```

```
## 0.3391011 0.4370029
```

```
## attr(,"credMass")
```

```
## [1] 0.95
```

## Results

Present your findings.

## 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.

Last, F., & Last, F. (2025). *Book Title*. Publisher.

Smith, A., & Johnson, C. (2023). *Title of the Online Article*. Retrieved from <https://www.example.com>.

## Appendix

### Code

```
# You can reference your code in the appendix (sample here).
```

### 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  $\pi$  in terms of  $\psi$ :

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

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