### SIMULATION RESULTS

We have here simulation results based on the JINI method. We compare with some other methods where possible.

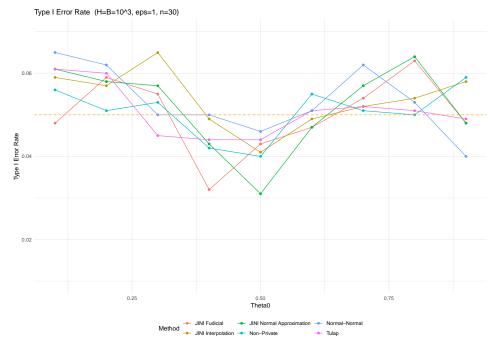
#### 1. Simple Hypothesis Testing

Given the observed sample  $X_1, \ldots, X_n \sim Ber(\theta_0)$ . We test the following hypothesis in differential privacy framework:

$$H_0: \theta_0 \le 0.9$$
  $vs$   $H_1: \theta_0 > 0.9$ .

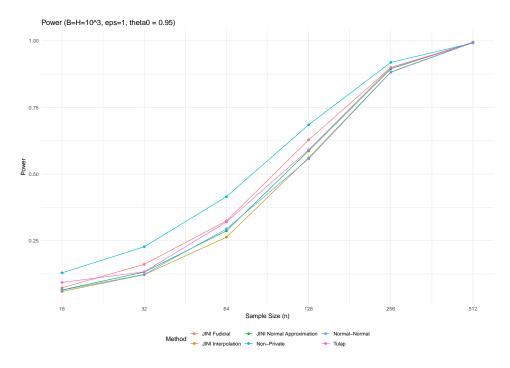
We study the power and level of the test. In what follows, B is the number of JINI bootstrap and H is the number of simulation.

1.1. Level Analysis. Simulation setting:  $n=30, B=10^3, H=10^3, \epsilon=1, \theta_0=0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9.$ 



### 1.2. Power Analysis. Simulation setting:

$$\theta_0 = 0.95, B = 10^3, H = 10^3, \epsilon = 1, n = 16, 32, 64, 128, 256, 512.$$



#### 2. Multiple Comparison

Here we consider comparing three groups. Given  $X_{1,j}, \ldots, X_{n,j} \sim Ber(\theta_{0_j})$ , j=1,2,3. Without loss of generality we set Group 1 as the reference group and define the log odd ratio of Groups 2 and 3 with respect to Group 1 as follows:

$$\beta_2 = \log\left(\frac{\theta_{0_2}}{1 - \theta_{0_2}}\right) - \log\left(\frac{\theta_{0_1}}{1 - \theta_{0_1}}\right)$$
$$\beta_3 = \log\left(\frac{\theta_{0_3}}{1 - \theta_{0_3}}\right) - \log\left(\frac{\theta_{0_1}}{1 - \theta_{0_1}}\right)$$

## 2.1. Case 1: We test the hypotheses:

$$H_0: \beta_i = 0$$
  $vs$   $H_1: \beta_i \neq 0, \quad i = 2, 3$   
Simulations Setting:  $\theta_{0_1} = 0.7, \theta_{0_2} = 0.5, \theta_{0_3} = 0.68; n_1 = 200, n_2 = 100, n_3 = 120; \alpha = 0.05, seed = 1234, B = 10^3, H = 10^3$ 

#### 3. Confidence Intervals

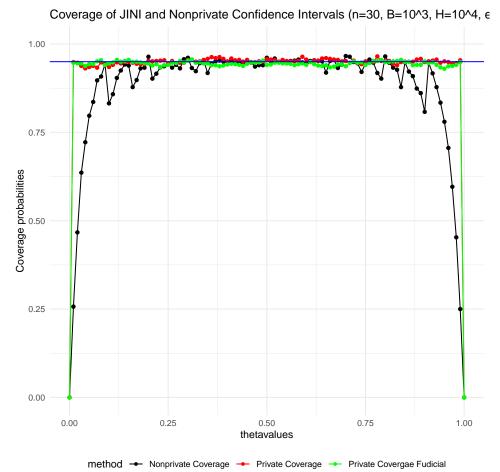
We use JINI method to build confidence interval for a population parameter in the DP framework.

Group	true $\beta$	est. $\hat{\beta}$	p-value	Reject H <sub>0</sub>
Group 2	-0.8473	-0.5585	0.0009	TRUE
Group 3	-0.0935 2	-0.5273	0.0006	TRUE

Table 1. Hypothesis testing.

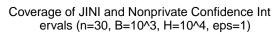
3.1. Coverage Analysis. Consider  $X_1, \ldots, X_n \sim Ber(\theta_0)$ . We study the coverage of the CI for  $\theta_0$  in DP setting.

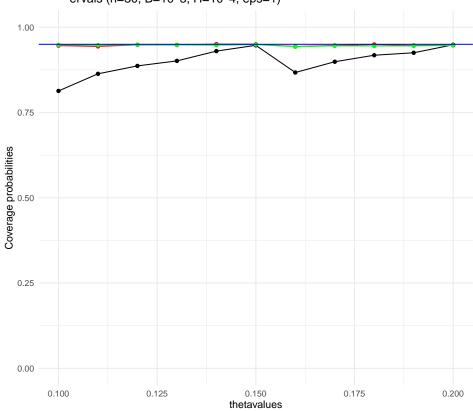
Simulation Setting:  $\theta_0 = 0.000.010.020.030.040.05, \dots, 1, n = 30, B = 10^3, H = 10^4, \epsilon = 1, \alpha = 0.05, seed = 345$ 



From the above coverage plot, the JINI private methods appear to be doing much better than the non-private method for values of  $\theta_0$  near 0 and 1.

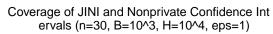
# 3.1.1. Coverage for $\theta_0$ near 0.

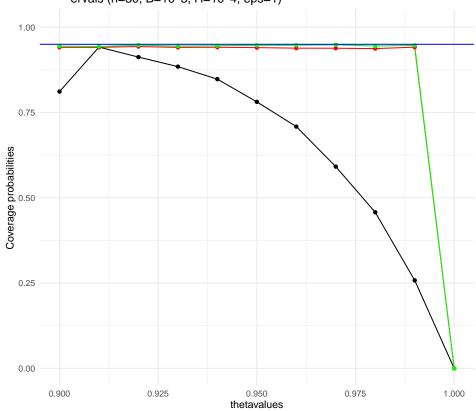




method → Nonprivate Coverage → Private Coverage → Private Covergae Fudicial

# 3.1.2. Coverage for $\theta_0$ near 1.

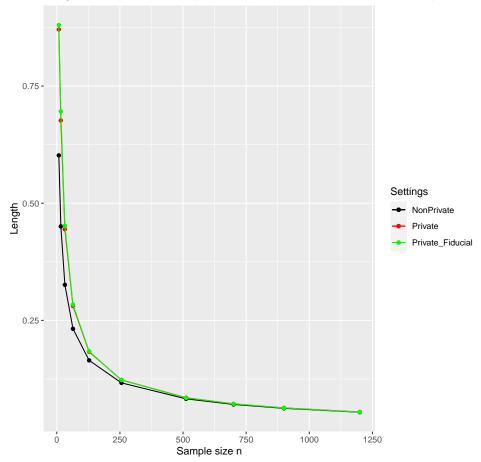




method → Nonprivate Coverage → Private Coverage → Private Covergae Fudicial

## 3.2. Length of Confidence Intervals.

Length of Private and Non-private CI (theta0=0.65, B=10^3, H=10^4, eps = 0.5)



3.3. Multiple Comparison. Here we are comparing three groups in terms of their log odd ratios. Precisely, we set Group 1 as reference and compare the other two groups with Group 1.

Given  $X_{1,j}, \ldots, X_{n,j} \sim Ber(\theta_{0_j})$ , j = 1, 2, 3. Define the log odd ratio of Groups 2 and 3 relative to Group 1 as follows:

$$\beta_2 = \log\left(\frac{\theta_{0_2}}{1 - \theta_{0_2}}\right) - \log\left(\frac{\theta_{0_1}}{1 - \theta_{0_1}}\right)$$

$$\beta_3 = \log\left(\frac{\theta_{0_3}}{1 - \theta_{0_3}}\right) - \log\left(\frac{\theta_{0_1}}{1 - \theta_{0_1}}\right)$$

We build confidence intervals for  $\beta_1$  and  $\beta_2$ .

