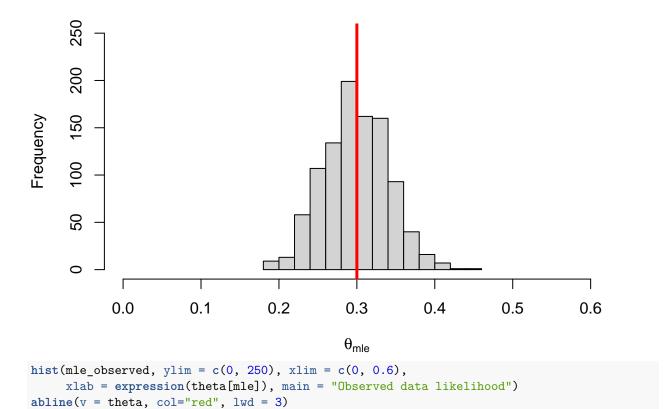
# University of Edinburgh, School of Mathematics Incomplete Data Analysis, 2020/2021 Ignorability – simulation study

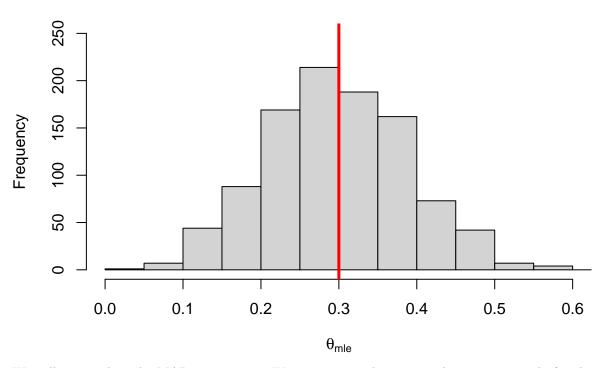
#### Vanda Inácio

In this supplementary file I show how to reproduce the results from the simulation study presented in the slides. We are assuming that  $Y_i \stackrel{\text{iid}}{\sim} \text{Bernoulli}(\theta)$  and  $R_i \mid Y_i \stackrel{\text{iid}}{\sim} \text{Bernoulli}(\theta)$ . I will simulate nsim = 1000 datasets of sample size n = 100 and consider  $\theta = 0.3$ . I will store the generated data and corresponding missing data indicators in a  $n \times nsim$  matrix. The maximum likelihood estimates (from the 1000 simulated datasets) based on both the full and observed data likelihood will be stored in two separate vectors.

## Full likelihood



### Observed data likelihood



We will now violate the MAR assumption. We repeat a similar exercise but now, instead of violating the

non-distinctness of parameters assumption, we violate the MAR assumption. In particular, we assume

$$Y_i \stackrel{\text{iid}}{\sim} \operatorname{Bernoulli}(\theta), \quad \text{and} \quad \Pr(R_i = 1 \mid Y_i) = \frac{e^{Y_i}}{1 + e^{Y_i}}.$$

The code follows below.

#### **MNAR** data

