

We derive the likelihood ratio statistics for testing whether the proportion of a given bin  $B$  has changed over-time.

$B$  can be, for example, the second bin,  $(0, 1000]$ .

There are 7 possible choices for  $B$  since 'min investment accepted' is discretized into 7 bins.

Notations:

- $t$  denotes the  $t^{\text{th}}$  quarter,  $1 \leq t \leq T$ . e.g.  $t = 2021Q4$ .
- $m_i$  is the 'minimum accepted investment' of the  $i^{\text{th}}$  observation
- $t_i$  is the quarter of 'sale date' of the  $i^{\text{th}}$  observation
- $p_{t_i}$  is the true probability of being in bin  $B$  in quarter  $t_i$

Let  $X_{it} = \begin{cases} 1, & \text{if } m_i \in B \\ 0, & \text{otherwise} \end{cases}$ . In other words,  $X_{it}$  indicates whether or not the minimum accepted investment of the  $i^{\text{th}}$  observation,  $m_i$ , is in bin  $B$ .

We assume that:

- 1) Each observation is independent
- 2)  $x_{it} \sim \text{Bernoulli}(p_{t,i})$

$$\text{So, } f(x_{it} | p_{t,i}) = p_{t,i}^{x_{it}} \cdot (1 - p_{t,i})^{1-x_{it}}$$

We form a hypothesis test:

$$H_0: p_1 = p_2 = \dots = p_T$$

$$H_1: \exists i, j \text{ s.t. } p_i \neq p_j.$$

Then,  $\theta = (\hat{p}_1, \hat{p}_2, \dots, \hat{p}_T)$ , given that there are  $m$  quarters,

Then, the overall likelihood is given by:

$$L(\theta) = \prod_{i=1}^n f(x_{it} | p_{t,i})$$

$$= \prod_{i=1}^n p_{t,i}^{x_{it}} \cdot (1 - p_{t,i})^{1-x_{it}}$$

$$= \prod_{t=1}^m \prod_{i \in S_t} p_t^{x_{it}} \cdot (1 - p_t)^{1-x_{it}}, \text{ where } S_t \text{ contain the}$$

observations in quarter  $t$

Under the null hypothesis,  $L(\theta) = \prod_{i=1}^n p^{x_i} (1-p)^{1-x_i}$ , so we are just maximizing one big joint-bernoullie likelihood, and we'll get  $\hat{p}_1 = \dots = \hat{p}_T = \frac{\sum x_i}{n} = \bar{x}$

Otherwise,  $L(\theta) = \prod_{t=1}^T \prod_{i \in S_t} p_t^{x_i} \cdot (1-p_t)^{1-x_i}$ , so we will maximize the joint-bernoullie likelihood within each quarter, and will get  $\hat{p}_1 = \frac{\sum_{i \in S_1} x_i}{n_1} = \bar{x}_1$

$$\hat{p}_2 = \frac{\sum_{i \in S_2} x_i}{n_2} = \bar{x}_2$$

⋮

$$\hat{p}_T = \frac{\sum_{i \in S_T} x_i}{n_T} = \bar{x}_T$$

Thus,  $\max_{\theta \in \theta_0} L(\theta) = \prod_{i=1}^n (\bar{x})^{x_i} \cdot (1 - \bar{x})^{1-x_i}$

$$\max_{\theta \notin \theta_0} L(\theta) = \prod_{t=1}^T \prod_{i \in S_t} (\bar{x}_t)^{x_i} \cdot (1 - \bar{x}_t)^{1-x_i}$$

and the test statistics is:

$$L = -2 \log \left( \frac{\max_{\theta \in \theta_0} L(\theta)}{\max_{\theta \notin \theta_0} L(\theta)} \right)$$

$$= -2 \left( \ln \bar{x} \cdot \sum_{i=1}^n x_i + \ln(1 - \bar{x}) \cdot \left( n - \sum_{i=1}^n x_i \right) - \sum_{t=1}^T \ln(\bar{x}_t) \cdot \sum_{i \in S_t} x_i + \ln(1 - \bar{x}_t) \cdot \left( n_t - \sum_{i \in S_t} x_i \right) \right)$$

where  $L \sim \chi^2_{m-1}$ .