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(1)

We observe X_i, Z_1, \ldots, Z_J , and d_i

The likelihood of choosing j and J respectively is:

$$P_{ij} = rac{\exp(u_{ij} - u_{iJ})}{1 + \sum_{k=1}^{J-1} \exp(u_{ik} - u_{iJ})}$$

$$P_{iJ} = \frac{1}{1 + \sum_{k=1}^{J-1} \exp(u_{ik} - u_{iJ})}$$

The log likelihood function we maximize is:

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