

Derivation of the Logit Probability

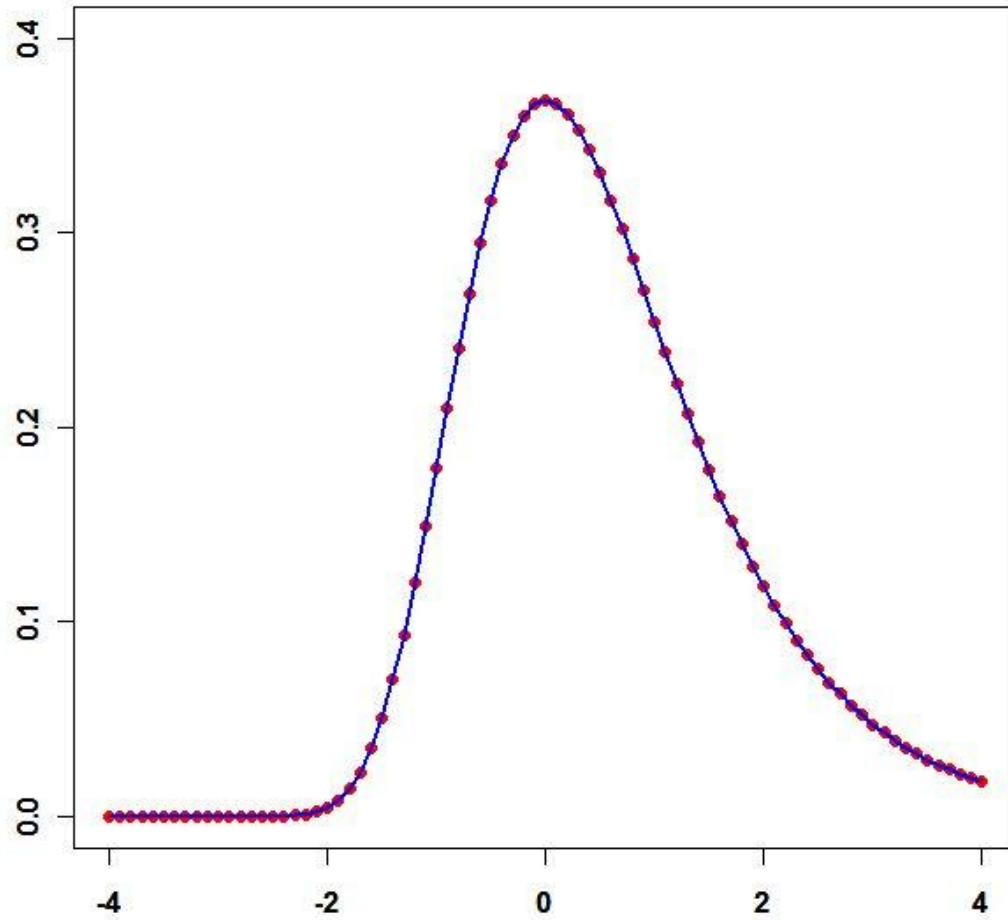
Utility function for Yea and Nay choices:

$$U_{iy} = e^{-d_{iy}^2} + \varepsilon_{iy} \quad \text{and} \quad U_{in} = e^{-d_{in}^2} + \varepsilon_{in} \quad (1)$$

where d_{iy}^2 and d_{in}^2 are the squared distances from the i th legislator to the Yea and Nay choices and the ε are distributed as the logarithm of the inverse of an exponential variable (Dhrymes, 1978, p. 342). Namely

$$f(\varepsilon) = e^{-\varepsilon} e^{-e^{-\varepsilon}}, \quad -\infty < \varepsilon < +\infty \quad (2)$$

The Log of the Inverse Exponential Distribution



The probability that the legislator will choose the Yea alternative is:

$$\mathbf{P}(U_{iy} > U_{in}) = \mathbf{P}(e^{-d_{iy}^2} + \varepsilon_{iy} > e^{-d_{in}^2} + \varepsilon_{in}) = \mathbf{P}(e^{-d_{iy}^2} - e^{-d_{in}^2} > \varepsilon_{in} - \varepsilon_{iy}) =$$

$$\mathbf{P}(\varepsilon_{in} - \varepsilon_{iy} < e^{-d_{iy}^2} - e^{-d_{in}^2}) = \mathbf{P}(\varepsilon_{iy} - \varepsilon_{in} > e^{-d_{in}^2} - e^{-d_{iy}^2}) \quad (3)$$

In order to get the distribution of $\boldsymbol{\varepsilon}_{iy} - \boldsymbol{\varepsilon}_{in}$ set up the joint density and then do a change of variables (note that the distribution of $\boldsymbol{\varepsilon}_{in} - \boldsymbol{\varepsilon}_{iy}$ will be the same as the distribution of $\boldsymbol{\varepsilon}_{iy} - \boldsymbol{\varepsilon}_{in}$):

$$\mathbf{f}(\boldsymbol{\varepsilon}_{iy}, \boldsymbol{\varepsilon}_{in}) = e^{-(\boldsymbol{\varepsilon}_{iy} + \boldsymbol{\varepsilon}_{in})} e^{-(e^{-\boldsymbol{\varepsilon}_{iy}} + e^{-\boldsymbol{\varepsilon}_{in}})} \quad (4)$$

Set $\mathbf{y} = \boldsymbol{\varepsilon}_{iy} - \boldsymbol{\varepsilon}_{in}$ and $\mathbf{z} = \boldsymbol{\varepsilon}_{in}$

Hence $\boldsymbol{\varepsilon}_{iy} = \mathbf{y} + \mathbf{z}$ and $\boldsymbol{\varepsilon}_{in} = \mathbf{z}$

and the Jacobian is:

$$\mathbf{J} = \begin{vmatrix} \frac{\partial \boldsymbol{\varepsilon}_{iy}}{\partial \mathbf{y}} & \frac{\partial \boldsymbol{\varepsilon}_{iy}}{\partial \mathbf{z}} \\ \frac{\partial \boldsymbol{\varepsilon}_{in}}{\partial \mathbf{y}} & \frac{\partial \boldsymbol{\varepsilon}_{in}}{\partial \mathbf{z}} \end{vmatrix} = \begin{vmatrix} \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{1} \end{vmatrix} = \mathbf{1}$$

Hence

$$\mathbf{f}(\mathbf{y} + \mathbf{z}, \mathbf{z}) = e^{-(\mathbf{y} + 2\mathbf{z})} e^{-(e^{-(\mathbf{y} + \mathbf{z})} + e^{-\mathbf{z}})} = e^{-\mathbf{y}} e^{-2\mathbf{z}} e^{-[e^{-\mathbf{z}}(1 + e^{-\mathbf{y}})]}$$

To get the distribution of $\mathbf{y} = \boldsymbol{\varepsilon}_{iy} - \boldsymbol{\varepsilon}_{in}$ integrate out \mathbf{z} :

$$\int_{-\infty}^{+\infty} e^{-\mathbf{y}} e^{-2\mathbf{z}} e^{-[e^{-\mathbf{z}}(1 + e^{-\mathbf{y}})]} d\mathbf{z} \quad (5)$$

This requires another change in variables:

Set $\mathbf{v} = e^{-\mathbf{z}}(1 + e^{-\mathbf{y}})$

Note that $\mathbf{0} < \mathbf{v} < \infty$ because $\mathbf{0} < e^{-\mathbf{z}} < \infty$ as $-\infty < \mathbf{z} < \infty$

Hence, $\ln(\mathbf{v}) = -\mathbf{z} + \ln(\mathbf{1} + \mathbf{e}^{-\mathbf{y}})$

and $\mathbf{z} = \ln(\mathbf{1} + \mathbf{e}^{-\mathbf{y}}) - \ln(\mathbf{v})$

Therefore, $\frac{\partial \mathbf{z}}{\partial \mathbf{v}} = -\frac{1}{\mathbf{v}}$ and $\mathbf{J} = \left| \frac{\partial \mathbf{z}}{\partial \mathbf{v}} \right| = \frac{1}{\mathbf{v}}$

Hence

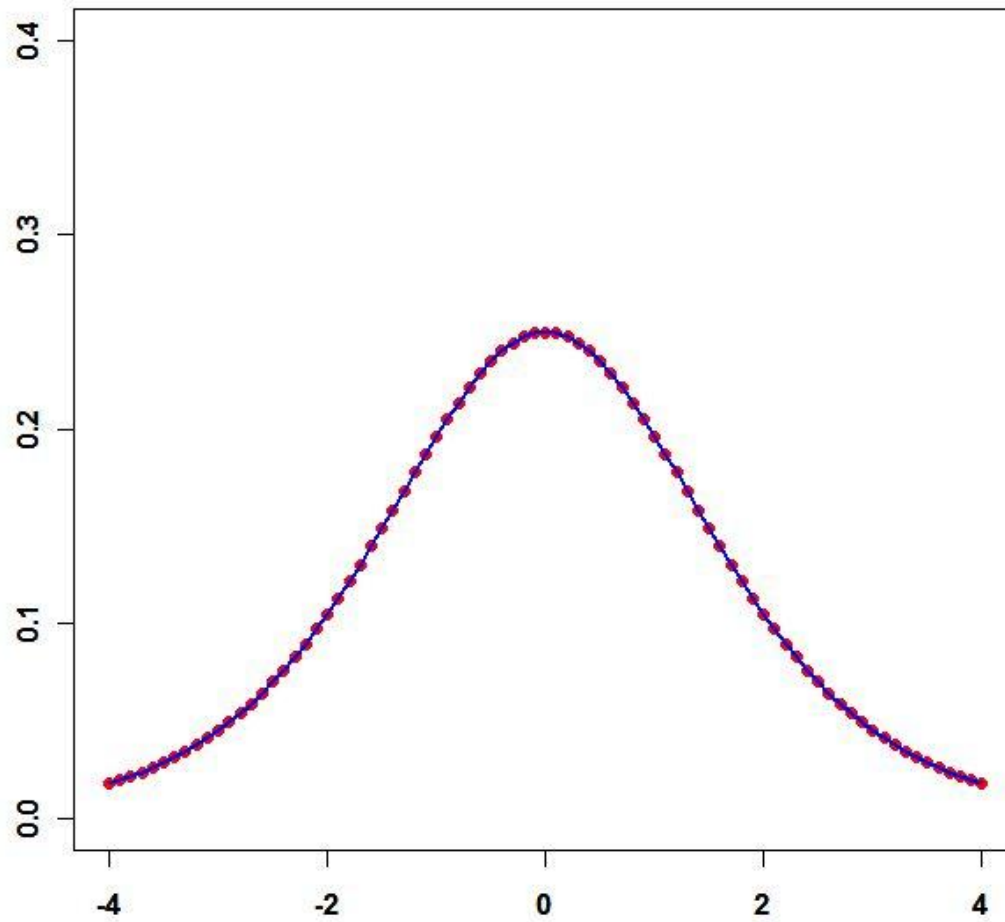
$$\int_{-\infty}^{+\infty} \mathbf{e}^{-\mathbf{y}} \mathbf{e}^{-2\mathbf{z}} \mathbf{e}^{-[\mathbf{e}^{-\mathbf{z}}(\mathbf{1} + \mathbf{e}^{-\mathbf{y}})]} \mathbf{d}\mathbf{z} = \mathbf{e}^{-\mathbf{y}} \int_0^{+\infty} \mathbf{e}^{-2\ln(\mathbf{1} + \mathbf{e}^{-\mathbf{y}})} \mathbf{e}^{2\ln(\mathbf{v})} \mathbf{e}^{-\{\mathbf{e}^{-[\ln(\mathbf{1} + \mathbf{e}^{-\mathbf{y}}) - \ln(\mathbf{v})]}[\mathbf{1} + \mathbf{e}^{-\mathbf{y}}]\}} \frac{1}{\mathbf{v}} \mathbf{d}\mathbf{v} =$$

$$\mathbf{e}^{-\mathbf{y}} \int_0^{+\infty} (\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-2} \mathbf{v}^2 \mathbf{e}^{-\{(\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-1} \mathbf{v}[\mathbf{1} + \mathbf{e}^{-\mathbf{y}}]\}} \frac{1}{\mathbf{v}} \mathbf{d}\mathbf{v} = \mathbf{e}^{-\mathbf{y}} (\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-2} \int_0^{+\infty} \mathbf{v}^2 \mathbf{e}^{-\mathbf{v}} \frac{1}{\mathbf{v}} \mathbf{d}\mathbf{v} =$$

$$\mathbf{e}^{-\mathbf{y}} (\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-2} \int_0^{+\infty} \mathbf{v} \mathbf{e}^{-\mathbf{v}} \mathbf{d}\mathbf{v} = \mathbf{e}^{-\mathbf{y}} (\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-2} \Gamma(2) = \mathbf{e}^{-\mathbf{y}} (\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^{-2} \quad (6)$$

$$\text{Therefore, } \mathbf{f}(\boldsymbol{\varepsilon}_{\text{iy}} - \boldsymbol{\varepsilon}_{\text{in}}) = \mathbf{f}(\mathbf{y}) = \frac{\mathbf{e}^{-\mathbf{y}}}{(\mathbf{1} + \mathbf{e}^{-\mathbf{y}})^2}, \quad (7)$$

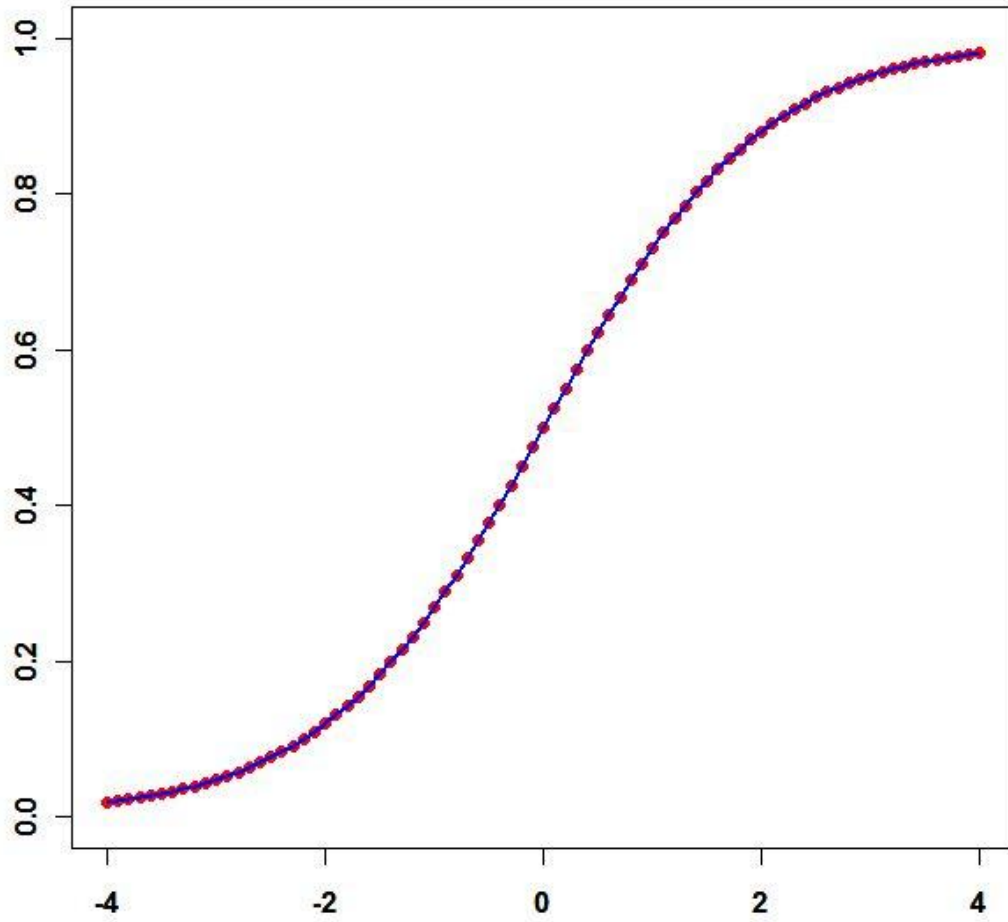
The Distribution of Two Random Draws From the Log of the Inverse Exponential Distribution



To get the distribution function:

$$F(y < t) = \int_{-\infty}^t \frac{e^{-y}}{(1+e^{-y})^2} dy = \left. \frac{t}{1+e^{-y}} \right|_{-\infty} = \frac{1}{1+e^{-t}}$$

Logit Probability: The Distribution Function of Two Random Draws From the Log of the Inverse Exponential Distribution



Hence

$$\mathbf{P}(\varepsilon_{in} - \varepsilon_{iy} < e^{-d_{iy}^2} - e^{-d_{in}^2}) = \frac{\mathbf{1}}{\mathbf{1} + e^{-(e^{-d_{iy}^2} - e^{-d_{in}^2})}} \quad \mathbf{(8)}$$