

Logistic Regression

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$$P(y = 1|x) = \mu(x) = \frac{1}{1 + e^{-\theta^T x}} = \frac{e^{x\theta}}{1 + e^{x\theta}}$$

Finding the Parameter, θ

$$X\theta = \log\left(\frac{P(Y|X)}{1 - P(Y|X)}\right)$$

- **Maximum Likelihood Estimation (MLE) of θ**

- Choose θ that maximizes the probability of observed data

$$\hat{\theta} = \operatorname{argmax}_{\theta} P(D|\theta)$$

- **This is Maximum Conditional Likelihood Estimation (MCLE)**

- $\hat{\theta} = \operatorname{argmax}_{\theta} P(D|\theta) = \operatorname{argmax}_{\theta} \prod_{1 \leq i \leq N} P(Y_i|X_i; \theta)$

$$= \operatorname{argmax}_{\theta} \log\left(\prod_{1 \leq i \leq N} P(Y_i|X_i; \theta)\right) = \operatorname{argmax}_{\theta} \sum_{1 \leq i \leq N} \log(P(Y_i|X_i; \theta))$$

- $P(Y_i|X_i; \theta) = \mu(X_i)^{Y_i} (1 - \mu(X_i))^{1-Y_i}$

- $\log(P(Y_i|X_i; \theta)) = Y_i \log(\mu(X_i)) + (1 - Y_i) \log(1 - \mu(X_i))$

$$= Y_i \{\log(\mu(X_i)) - \log(1 - \mu(X_i))\} + \log(1 - \mu(X_i))$$

$$= Y_i \log\left(\frac{\mu(X_i)}{1 - \mu(X_i)}\right) + \log(1 - \mu(X_i))$$

$$= Y_i X_i \theta + \log(1 - \mu(X_i)) = Y_i X_i \theta - \log(1 + e^{X_i \theta})$$

Finding the Parameter, θ , contd.

Linear Regression (Closed Form):

$$\begin{aligned}\hat{f} &= X\theta & \nabla_{\theta}(\theta^T X^T X \theta - 2\theta^T X^T Y) &= 0 \\ & & 2X^T X \theta - 2X^T Y &= 0 \\ & & \theta &= (X^T X)^{-1} X^T Y\end{aligned}$$

- $\hat{\theta} = \operatorname{argmax}_{\theta} \sum_{1 \leq i \leq N} \log(P(Y_i | X_i; \theta))$
- $= \operatorname{argmax}_{\theta} \sum_{1 \leq i \leq N} \{Y_i X_i \theta - \log(1 + e^{X_i \theta})\}$
- Partial derivative to find a certain element in θ

$$\frac{\partial}{\partial \theta_j} \left\{ \sum_{1 \leq i \leq N} Y_i X_i \theta - \log(1 + e^{X_i \theta}) \right\}$$

$$P(y = 1 | x) = \frac{e^{x\theta}}{1 + e^{x\theta}}$$

$$= \left\{ \sum_{1 \leq i \leq N} Y_i X_{i,j} \right\} + \left\{ \sum_{1 \leq i \leq N} -\frac{1}{1 + e^{X_i \theta}} \times e^{X_i \theta} \times X_{i,j} \right\}$$

$$= \sum_{1 \leq i \leq N} X_{i,j} \left(Y_i - \frac{e^{X_i \theta}}{1 + e^{X_i \theta}} \right) = \sum_{1 \leq i \leq N} X_{i,j} (Y_i - P(Y_i = 1 | X_i; \theta)) = 0$$

- There is no way to derive further
 - There is no closed form solution!
 - Open form solution \rightarrow approximate!

Cannot be easily solved in the closed form because of the logistic function