1 Zadanie pierwsze

Logistic model:

$$P(y = 1|x, \theta) = \sigma(x\theta) = \frac{1}{1 + e^{-x\theta}} = \frac{1}{1 + exp(-\sum_{i=1}^{5} x_i \theta_i + \theta_0)}$$

Bayes model:

$$P(y=1|x) = \frac{P(x|y=1)P(y=1)}{P(x)} = \frac{P(x|y=1)P(y=1)}{P(y=1)P(x|y=1) + P(y=0)P(x|y=0)} = \frac{P(x|y=1)P(y=1)}{P(x|y=1)P(x|y=1) + P(y=1)P(x|y=0)} = \frac{P(x|y=1)P(y=1)}{P(x|y=1)P(x|y=1) + P(y=1)P(x|y=0)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|y=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|x=1)P(x|y=1)}{P(x|x=1)P(x|y=1)} = \frac{P(x|x=1)P(x|y=1)}{P(x|x=1)P(x|x=1)} = \frac{P(x|x=1)P(x|x=1)}{P(x|x=1)P(x|x=1)} = \frac{P($$

After dividing by numerator

$$=\frac{1}{1+\frac{P(x|y=0)P(y=0)}{P(x|y=1)P(y=1)}}=\frac{1}{1+exp(ln(\frac{P(x|y=0)P(y=0)}{P(x|y=1)P(y=1)}))}=\frac{1}{1+exp(ln(\frac{P(x|y=0)}{P(x|y=1)})+ln(\frac{P(y=0)}{P(y=1)}))}$$

Naive assumption:

$$P(x|y=0) = \prod_{i=1}^{5} P(x_i|y=0)$$

$$P(x|y=1) = \prod_{i=1}^{5} P(x_i|y=1)$$

Rewriting previous expression we get:

$$P(y=1|x) = \frac{1}{1 + exp(ln\frac{P(y=0)}{P(y=1)} + \sum ln\frac{x_i|y=0)}{P(x_i|y=1)})}$$

Therefore, logistic regression is equivalent to naive Bayes when:

$$\theta_0 = ln \frac{P(y=0)}{P(y=1)}$$

$$-\sum_{i=1}^{5} x_i \theta_i = \sum \ln \frac{P(x_i|y=0)}{P(x_i|y=1)}$$