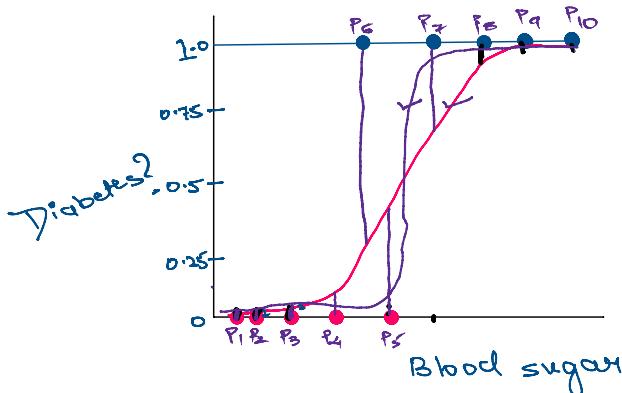


Logistic Regression 2

14 July 2024 11:51



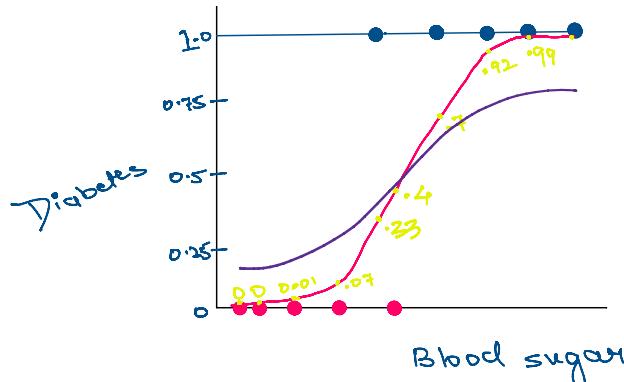
$$z = \beta_0 + \beta_1 x$$

$$\frac{1}{1 + e^{-z}}$$

$$\frac{1}{1 + e^{-(\beta_0 + \beta_1 x)}}$$

$$\text{Probability of Diabetes} = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x)}}$$

By changing the β_0 & β_1 , the curve changes.

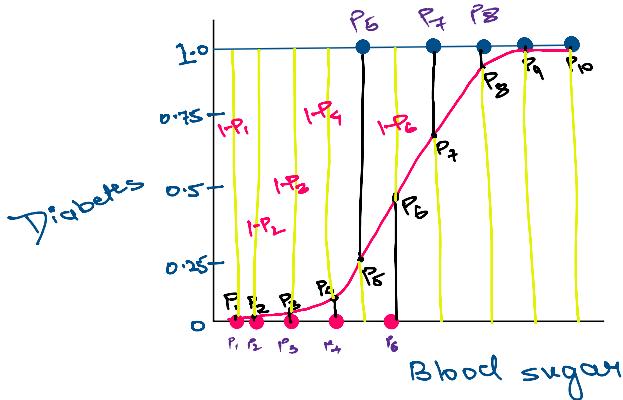


where suppose

$$\beta_0 = -15 \quad \text{and}$$

$$\beta_1 = 0.065$$

Finding the best fit Sigmoid curve



maximizes $(P_6 \times P_7 \times P_8 \times P_9 \times P_{10})$

minimizes $(P_1 \times P_2 \times P_3 \times P_4 \times P_5)$

So the best fitting combination of β_0 & β_1 would be the one where we maximize the product :

$$(1-P_1)(1-P_2)(1-P_3)(1-P_4)(1-P_5)(P_6)(P_7)(P_8)(P_9)(P_{10})$$

formula.

$$(1-P_1)(1-P_2)(1-P_3)(1-P_4)(1-P_5)(P_6)(P_7)(P_8)(P_9)(P_{10})$$

This product is called 'likelihood function'.

It is the product of:

$$[(1-P_i)(1-P_i) \dots] * [(P_i)(P_i) \dots]$$

\downarrow \downarrow

for all non-diabetics for all diabetics

MSE

maximizes $(P_5 \times P_7 \times P_8 \times P_9 \times P_{10})$

minimizes $(P_1 P_2 P_3 P_4 P_6)$

$$P_1 = 0.3 \quad 1 - P_1 = 0.7$$

$$P_1 = 0.05 \quad 1 - P_1 = 0.95$$

maximizes $(1 - P_1 \times 1 - P_2 \times 1 - P_3 \times 1 - P_4 \times 1 - P_6 \times P_5 \times P_7 \times P_8 \times P_9 \times P_{10})$

\uparrow