Maximum likelihood in logistic regression

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Goal

This document is help you understand how people came up the cost function of logistic regression:

$$J(heta) = -rac{1}{m}\sum_{i=1}^{m}\left[y^{(i)}\logig(h_{ heta}\left(x^{(i)}
ight)ig) + ig(1-y^{(i)}ig)\logig(1-ig(h_{ heta}\left(x^{(i)}ig)ig)
ight]$$

Function set

Imagine a 2-class problem. C1 and C2

The purpose of logistic regression is that we want to find the probability of which class it should be for a given x.

So for a given x, we want to find $P_{w,b}$ $(C_1 \mid x)$, if $P_{w,b}$ $(C_1 \mid x) > 0.5$, the result is C1, otherwise, output C2 .

A good way to change output to 0~1 range is sigmoid function.

$$P_{w,b}\left(C_1\mid x
ight) = \sigma(z) \ \sigma(z) = rac{1}{1+\exp(-z)}$$

$$z = w \cdot x + b = \sum_i w_i x_i + b$$

Now you got your function $f_{w,b}(x) = \sigma\left(\sum_i w_i x_i + b\right)$ whose output is between 0 and 1

Likelihood of the function

From the function set we could see that $f_{w,b}(x) = P_{w,b}\left(C_1 \mid x\right)$

And we got a lot of training data

$$x^1 \quad x^2 \quad x^3 \quad \cdots \quad x^N$$

$$C_1 \quad C_1 \quad C_2 \quad \cdots \quad C_1$$

Assume there is a real probability in this world, and our training data is generated from this probability $P_{w,b}\left(C_1\mid x\right)$

So what is the probability of generating this whole training dataset?

$$L(w,b) = P_{w,b} (C_1 \mid x^1) P_{w,b} (C_1 \mid x^2) (1 - P_{w,b} (C_1 \mid x^3)) \cdots (P_{w,b} (C_1 \mid x^N))$$

= $f_{w,b} (x^1) f_{w,b} (x^2) (1 - f_{w,b} (x^3)) \cdots f_{w,b} (x^N)$

So the most likely w* and b* is the one with the largest L(w,b)

$$w^*, b^* = \arg\max_{w,b} L(w, b)$$

Multiply operation is easy to cause overflow so we write this into log liklihood

$$w^*, b^* = \arg\min_{w,b} - \ln L(w,b)$$

$$egin{aligned} &-\ln L(w,b) = \ln f_{w,b}\left(x^1
ight) + \ln f_{w,b}\left(x^2
ight) + \ln \left(1 - f_{w,b}\left(x^3
ight)
ight) \cdots \ \hat{y}^n: 1 ext{ for class } 1,0 ext{ for class } 2 \ &= \sum_n - \left[\hat{y}^n \ln f_{w,b}\left(x^n
ight) + \left(1 - \hat{y}^n
ight) \ln (1 - f_{w,b}\left(x^n
ight))
ight] \end{aligned}$$