Computational Linguistics

Natural Language Processing

University of Maryland

Classification Examples

Reminder: Logistic Regression

$$P(Y=0|X) = \frac{1}{1 + \exp[\beta_0 + \sum_i \beta_i X_i]}$$
(1)
$$P(Y=1|X) = \frac{\exp[\beta_0 + \sum_i \beta_i X_i]}{1 + \exp[\beta_0 + \sum_i \beta_i X_i]}$$
(2)

- Discriminative prediction: p(y|x)
- Classification uses: ad placement, spam detection
- What we didn't talk about is how to learn β from data

Logistic Regression: Objective Function

$$\mathscr{L} \equiv \ln p(Y|X,\beta) = \sum_{j} \ln p(y^{(j)}|x^{(j)},\beta)$$

$$= \sum_{j} y^{(j)} \left(\beta_0 + \sum_{i} \beta_i x_i^{(j)}\right) - \ln \left[1 + \exp\left(\beta_0 + \sum_{i} \beta_i x_i^{(j)}\right)\right]$$
(3)
(4)

Algorithm

- 1. Initialize a vector B to be all zeros
- 2. For *t* = 1,..., *T*
 - For each example \vec{x}_i , y_i and feature *j*:
 - Compute $\pi_i \equiv \Pr(y_i = 1 | \vec{x}_i)$
 - Set $\beta[j] = \beta[j]' + \lambda(y_i \pi_i)x_{i,j}$
- 3. Output the parameters β_1, \ldots, β_d .

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$

$$\vec{\beta} = \langle \beta_{bias} = 0, \beta_A = 0, \beta_B = 0, \beta_C = 0, \beta_D = 0 \rangle$$

$y_1 = 1$

A A A A B B B C (Assume step size $\lambda = 1.0.$)

$$y_2 = 0$$

BCCCDDDD

You first see the positive example. First, compute π_1

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
$$\vec{\beta} = \langle 0, 0, 0, 0, 0 \rangle$$

 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

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You first see the positive example. First, compute π_1 $\pi_1 = \Pr(y_1 = 1 | \vec{x_1}) = \frac{\exp \beta^T x_i}{1 + \exp \beta^T x_i} = \frac{\exp 0}{\exp 0 + 1} = 0.5$

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
$$\vec{\beta} = \langle 0, 0, 0, 0, 0 \rangle$$

*y*₁ =1

A A A A B B B C (Assume step size $\lambda = 1.0.$)

 $\pi_1 = 0.5$ What's the update for β_{bias} ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
$$\vec{\beta} = \langle 0, 0, 0, 0, 0 \rangle$$

 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

What's the update for β_{bias} ? $\beta_{bias} = \beta'_{bias} + \lambda \cdot (y_1 - \pi_1) \cdot x_{1,bias} = 0.0 + 1.0 \cdot (1.0 - 0.5) \cdot 1.0$

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$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$

A A A A B B B C (Assume step size $\lambda = 1.0.$) *y*₂ =0 B C C C D D D D

What's the update for β_A ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

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A A A A B B B C (Assume step size $\lambda = 1.0.$) *y*₂ =0 B C C C D D D D

What's the update for β_B ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

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A A A A B B B C (Assume step size $\lambda = 1.0.$) *y*₂ =0 B C C C D D D D

What's the update for β_C ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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A A A A B B B C (Assume step size $\lambda = 1.0.$) *y*₂ =0 B C C C D D D D

What's the update for β_D ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
$$\vec{\beta} = \langle 0, 0, 0, 0, 0 \rangle$$

 $y_1 = 1$

A A A A B B B C (Assume step size $\lambda = 1.0.$) $y_2 = 0$ BCCCDDDD

What's the update for β_D ? $\beta_D = \beta'_D + \lambda \cdot (y_1 - \pi_1) \cdot x_{1,D} = 0.0 + 1.0 \cdot (1.0 - 0.5) \cdot 0.0$

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
$$\vec{\beta} = \langle .5, 2, 1.5, 0.5, 0 \rangle$$

 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

Now you see the negative example. What's π_2 ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

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$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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<i>y</i> ₁ =1	<i>y</i> ₂ =0
АААВВВС	BCCCDDDD
(Assume step size $\lambda =$ 1.0.)	

Now you see the negative example. What's π_2 ? $\pi_2 = 0.97$ What's the update for β_{bias} ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$ A A A A B B B C (Assume step size $\lambda = 1.0.$) $y_2 = 0$ B C C C D D D D

What's the update for β_{bias} ? $\beta_{bias} = \beta'_{bias} + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,bias} = 0.5 + 1.0 \cdot (0.0 - 0.97) \cdot 1.0$

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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

What's the update for β_{bias} ? $\beta_{bias} = \beta'_{bias} + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,bias} = 0.5 + 1.0 \cdot (0.0 - 0.97) \cdot 1.0 = -0.47$

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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*y*₁ = 1

A A A A B B B C (Assume step size $\lambda = 1.0.$) $y_2 = 0$ BCCCDDDD

What's the update for β_A ?

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

What's the update for β_A ? $\beta_A = \beta'_A + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,A} = 2.0 + 1.0 \cdot (0.0 - 0.97) \cdot 0.0$

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What's the update for β_B ? $\beta_B = \beta'_B + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,B} = 1.5 + 1.0 \cdot (0.0 - 0.97) \cdot 1.0$

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 $y_1 = 1$ $y_2 = 0$ A A A A B B B CB C C C D D D D(Assume step size $\lambda = 1.0.$)

What's the update for β_B ? $\beta_B = \beta'_B + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,B} = 1.5 + 1.0 \cdot (0.0 - 0.97) \cdot 1.0 = 0.53$

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*y*₁ = 1

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What's the update for β_C ? $\beta_C = \beta'_C + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,C} = 0.5 + 1.0 \cdot (0.0 - 0.97) \cdot 3.0 = -2.41$

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$
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*y*₁ = 1

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What's the update for β_D ?

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What's the update for β_D ? $\beta_D = \beta'_D + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,D} = 0.0 + 1.0 \cdot (0.0 - 0.97) \cdot 4.0$

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What's the update for β_D ? $\beta_D = \beta'_D + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,D} = 0.0 + 1.0 \cdot (0.0 - 0.97) \cdot 4.0 = -3.88$

Different Activation Function

Your boss demands that you replace the sigmoid function in logistic regression with the trigonometric sin function because it looks the same and he has a sin button on his calculator.

- 1. Plot both between -1 and 1. What would a choice of constants *A* and *B* be that would make $s(z) = A\sin(Bz) + C$ look as much like the logistic function?
- 2. What would be the update for an example? How is it different?
- 3. Would there be any other problems with using this formulation?

- Same min and max
- Derivative at 0 should match
- Same value at 0

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- Derivative at 0 should match
- Same value at 0

$$s(z) \equiv \frac{1}{2}\sin\left(\frac{z}{2}\right) + \frac{1}{2} \tag{5}$$

- Same min and max : $s(\pi) = \frac{1}{2}sin(\frac{\pi}{2}) + \frac{1}{2} = \frac{1}{2} + \frac{1}{2} = 1$
- Derivative at 0 should match
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- Same min and max : $s(\pi) = \frac{1}{2} \sin(\frac{\pi}{2}) + \frac{1}{2} = \frac{1}{2} + \frac{1}{2} = 1$
- Derivative at 0 should match
- Same value at 0 : $s(0) = \frac{1}{2}\sin 0 + \frac{1}{2} = 0 + \frac{1}{2} = \frac{1}{2}$

$$s(z) \equiv \frac{1}{2}\sin\left(\frac{z}{2}\right) + \frac{1}{2} \tag{5}$$

- Same min and max : $s(\pi) = \frac{1}{2}sin(\frac{\pi}{2}) + \frac{1}{2} = \frac{1}{2} + \frac{1}{2} = 1$
- Derivative at 0 should match : $s'(0) = \frac{1}{4} \cos 0 = \frac{1}{4}$
- Same value at 0 : $s(0) = \frac{1}{2}\sin 0 + \frac{1}{2} = 0 + \frac{1}{2} = \frac{1}{2}$

$$s(z) \equiv \frac{1}{2}\sin\left(\frac{z}{2}\right) + \frac{1}{2} \tag{5}$$

Plot



Perfect match!

What can go wrong?

Can you think of two different examples that have the same update but shouldn't?

Can you think of two different examples that have the same update but shouldn't?

The update becomes a function of sine and cosine.

If the input the function is near zero, everything is fine, but beyond that, as the gradient is a periodic function (every $\frac{2\pi}{\beta}$ you'll repeat), the gradient is possible to be zero or inf.

If you are interested ...

Let us define:

$$\pi_i = \frac{1}{2} \sin\left(\frac{\beta^T x_i}{2}\right) + \frac{1}{2}$$

Thus,

$$\frac{\partial \pi_i}{\partial \beta_j} = \frac{1}{4} \cos\left(\frac{\beta^T x_i}{2}\right) x_{i,j} \tag{6}$$

If you are interested ...

To ease notation, let us further define:

$$z'=\frac{\beta^T x_i}{2}$$

Thus,

$$\frac{\partial \mathscr{L}_i}{\partial \beta_j} = \begin{cases} \frac{1}{\pi_i} \frac{\partial \pi_i}{\partial \beta_j} & \text{if } y_i = 1\\ \frac{1}{1-\pi_i} \left(-\frac{\partial \pi_i}{\partial \beta_j}\right) & \text{if } y_i = 0 \end{cases} = \begin{cases} \frac{\cos z'}{2(\sin z'+1)} x_{i,j} & \text{if } y_i = 1\\ \frac{\cos z'}{2(\sin z'-1)} x_{i,j} & \text{if } y_i = 0 \end{cases}$$
(7)