



# **Logistic Regression**

Introduction to Data Science Algorithms
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ABC

## Logistic Regression: Objective Function

$$\ell \equiv \ln \rho(Y|X,\beta) = \sum_{j} \ln \rho(y^{(j)}|x^{(j)},\beta)$$

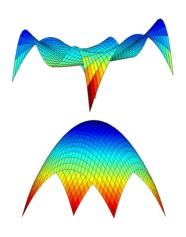
$$= \sum_{j} y^{(j)} \left(\beta_0 + \sum_{i} \beta_i x_i^{(j)}\right) - \ln \left[1 + \exp\left(\beta_0 + \sum_{i} \beta_i x_i^{(j)}\right)\right]$$
(2)

$$\ell \equiv \ln \rho(Y|X,\beta) = \sum_{j} \ln \rho(y^{(j)}|x^{(j)},\beta)$$

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(2)

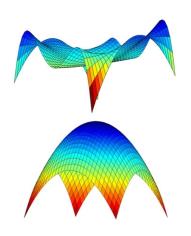
Training data (y, x) are fixed. Objective function is a function of  $\beta$  ... what values of  $\beta$  give a good value.

#### Convexity



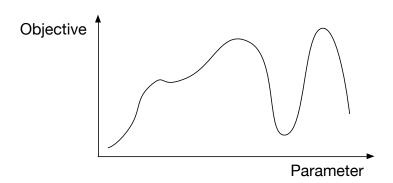
- Convex function
- Doesn't matter where you start, if you "walk up" objective

## Convexity

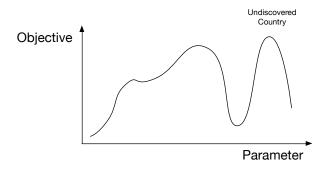


- Convex function
- Doesn't matter where you start, if you "walk up" objective
- Gradient!

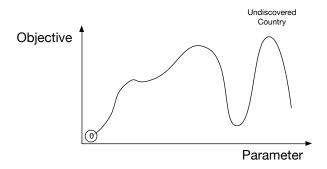
## Goal



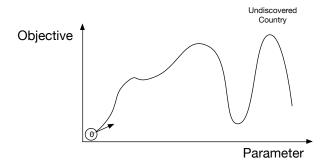
## Goal



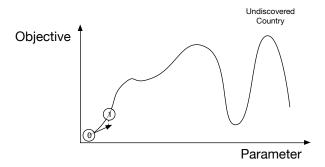
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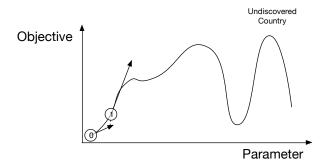
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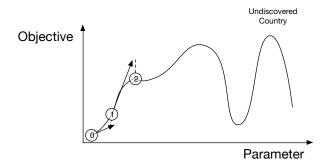
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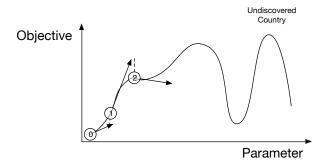
## Goal



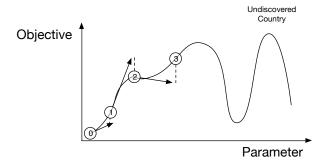
## Goal



## Goal

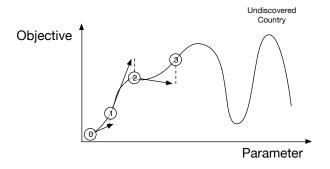


#### Goal



#### Goal

Optimize log likelihood with respect to variables eta



Luckily, (vanilla) logistic regression is convex