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# Lecture Notes for Machine Learning in Python

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Professor Eric Larson  
Week Five, Lecture B

# Class Logistics and Agenda

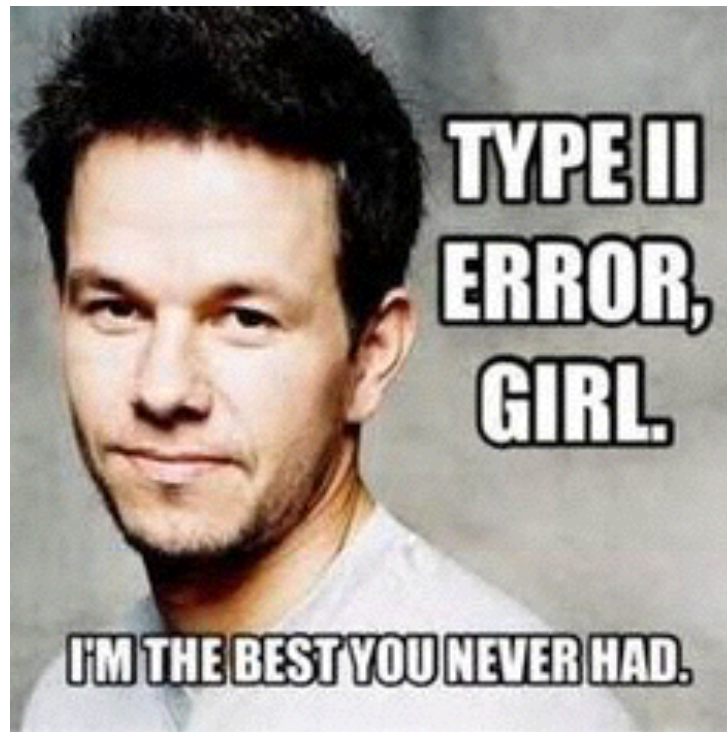
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- Welcome back to lecture!
- Agenda
  - Logistic Regression
    - Solving
    - Programming

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# Logistic Regression

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# Setting Up Binary Logistic Regression

- From flipped lecture:

$$\hat{y} = w^T \hat{x}$$

$$p(y^{(i)}=1 | x^{(i)}, w) = \frac{1}{1 + \exp(-w^T x^{(i)})}$$

$$p(y^{(i)}=0 | x^{(i)}, w) = 1 - \frac{1}{1 + \exp(-w^T x^{(i)})}$$

$$L(w) = \prod_{y^{(i)}=1} p(y^{(i)}=1 | x^{(i)}, w) \prod_{y^{(i)}=0} p(y^{(i)}=0 | x^{(i)}, w)$$

$$\text{MAX } L(w)$$

$$w^* = \underset{w}{\text{ARGMAX}} L(w)$$

maximize!

# Binary Solution for Update Equation

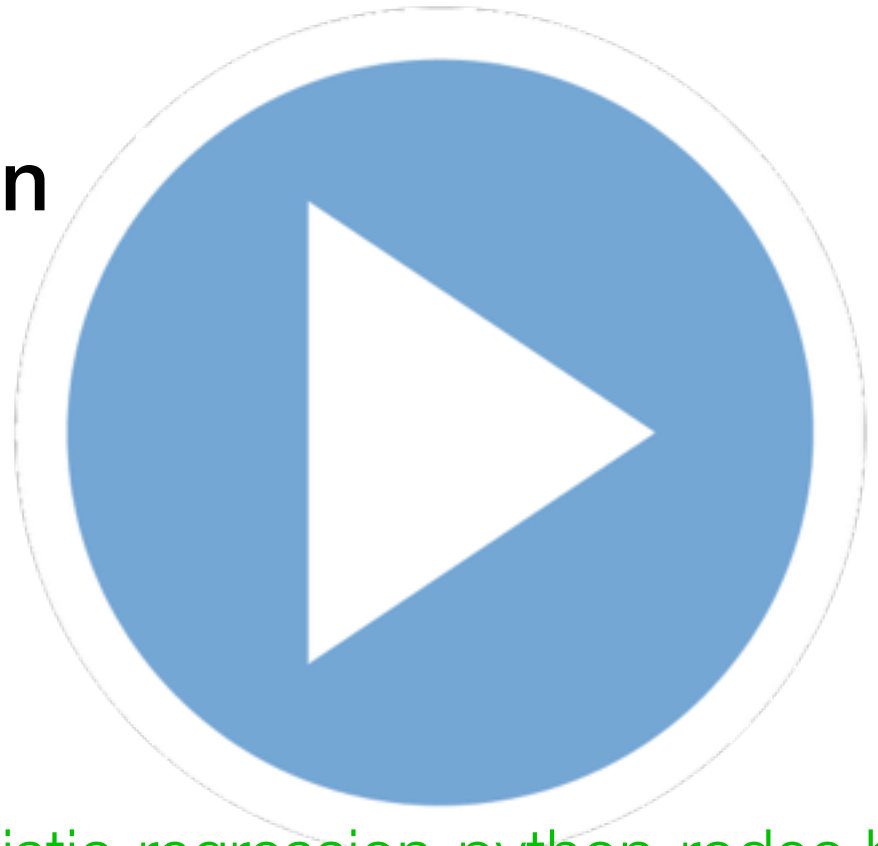
- From supplement:

$$\underbrace{w_j}_{\text{new value}} \leftarrow \underbrace{w_j}_{\text{old value}} + \underbrace{\eta \sum_{i=1}^M (y^{(i)} - g(x^{(i)})) x_j^{(i)}}_{\text{gradient}}$$

$$w \leftarrow w + \eta \sum_{i=1}^M (y^{(i)} - g(x^{(i)})) x^{(i)}$$

## *Reinvent sklearn* Logistic Regression

Programming  
Extended Example



## Other Tutorials:

<http://blog.yhat.com/posts/logistic-regression-python-rodeo.html>

[http://scikit-learn.org/stable/auto\\_examples/linear\\_model/plot\\_iris\\_logistic.html](http://scikit-learn.org/stable/auto_examples/linear_model/plot_iris_logistic.html)

# For Next Lecture

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- **Next time:** Gradient based optimization
- **Next Next time:** SVMs via in class assignment

# Scratch Paper



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