

# L2 Regularization in Logistic Regression

Classification Methods

Daniel E. Acuna

Associate Professor, University of Colorado Boulder



# Contents of This Video

In this video, we will cover:

- The overfitting problem in student classification
- What is L2 regularization?
- Mathematical formulation of regularized loss
- Effect of regularization parameter  $\lambda$  (lambda)
- Visualizing smooth vs. complex decision boundaries



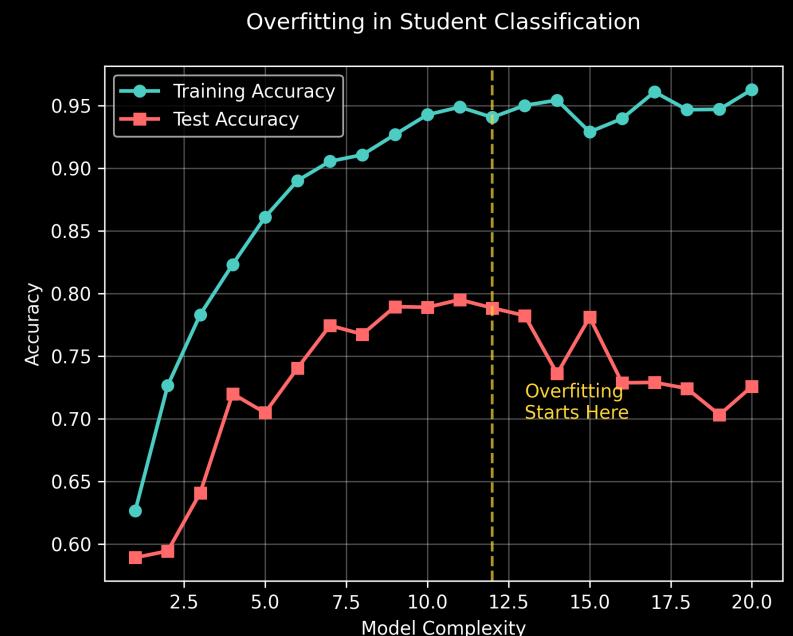
# The Overfitting Problem

## Student Success Prediction Challenge:

- Many features: study hours, attendance, assignments, GPA, sleep, job status, etc.
- Complex patterns in training data
- Model might memorize noise instead of true patterns

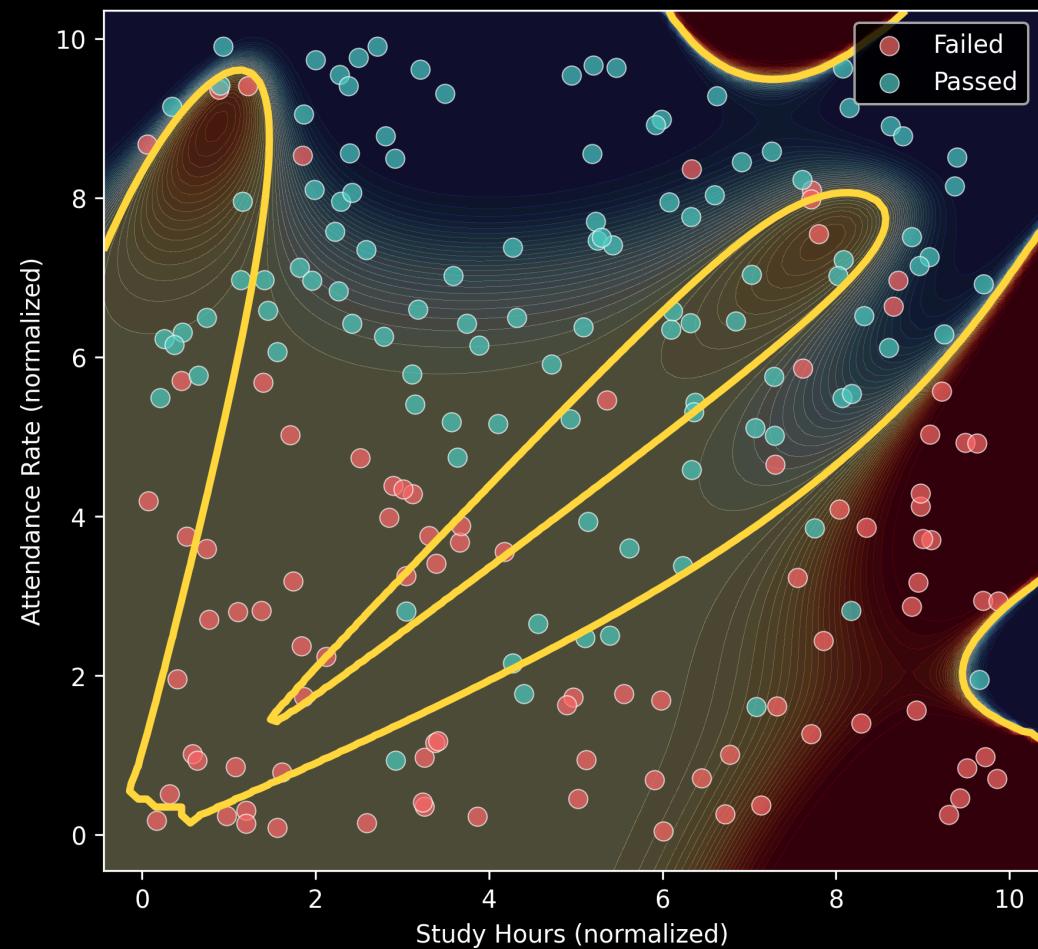
## Overfitting Signs:

- Perfect accuracy on training students
- Poor performance on new students
- Overly complex decision boundaries

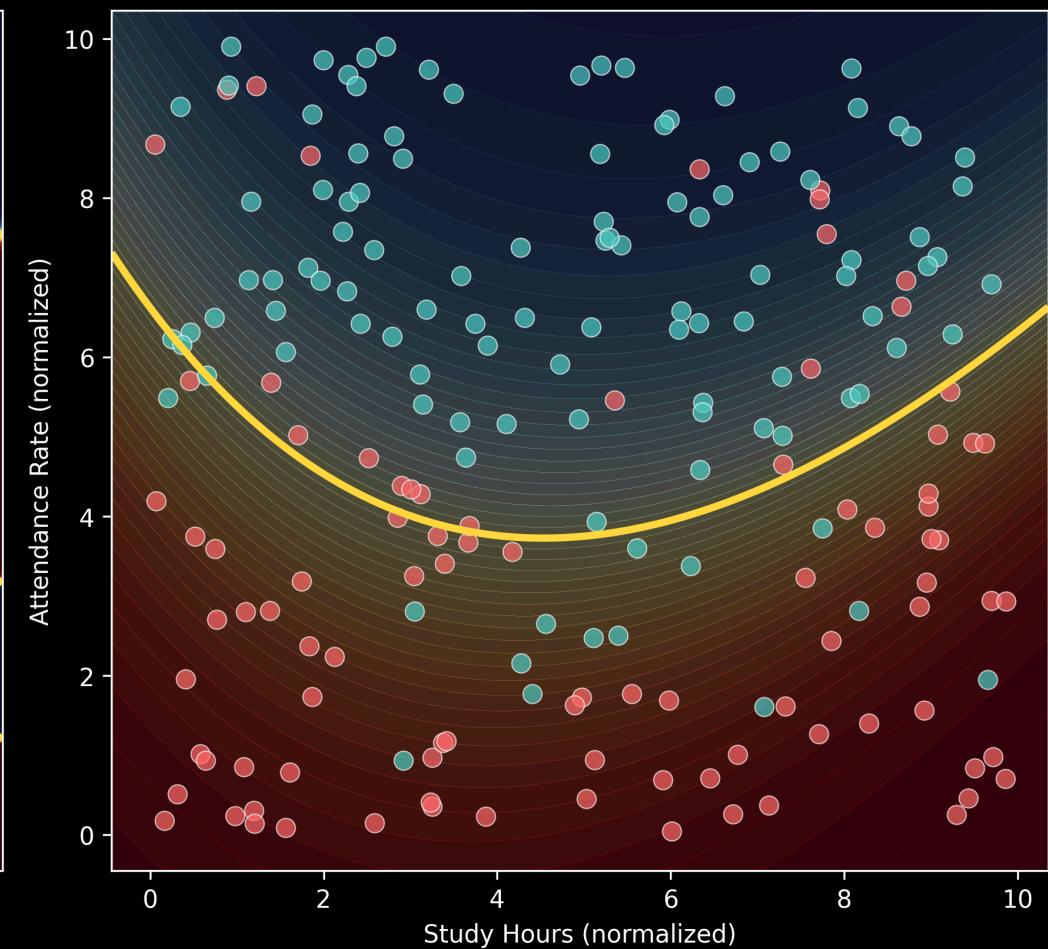


# Complex vs. Simple Decision Boundaries

Overfit (Degree 15, C=1000)



Regularized (Degree 2, C=0.1)



# L2 Regularization Mathematical Formulation

Original Loss Function:

$$J(\mathbf{w}, b) = -\frac{1}{m} \sum_{i=1}^m [y^{(i)} \log(y^{(i)}) + (1 - y^{(i)}) \log(1 - y^{(i)})]$$

Regularized Loss Function:

$$J_{reg}(\mathbf{w}, b) = J(\mathbf{w}, b) + \lambda \sum_{j=1}^p w_j^2$$

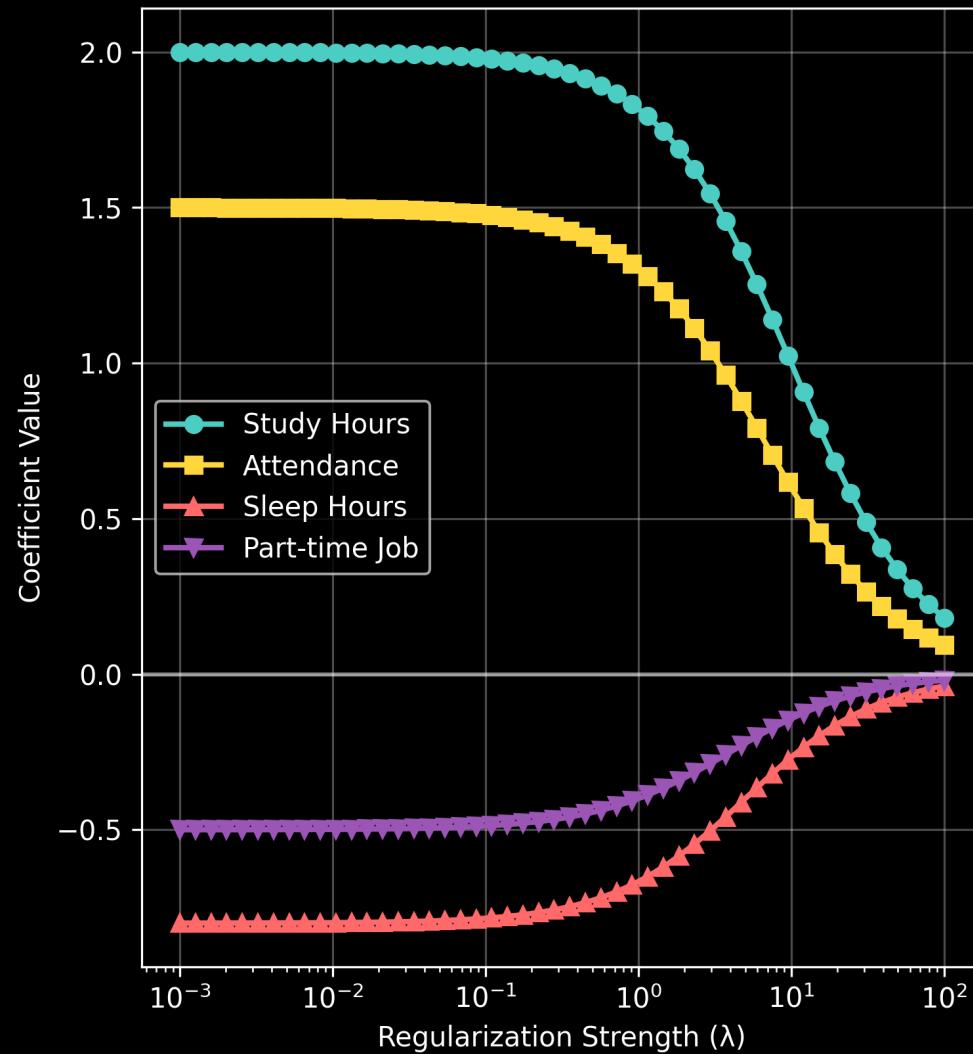
Where:

- $\lambda$  (lambda) = regularization strength

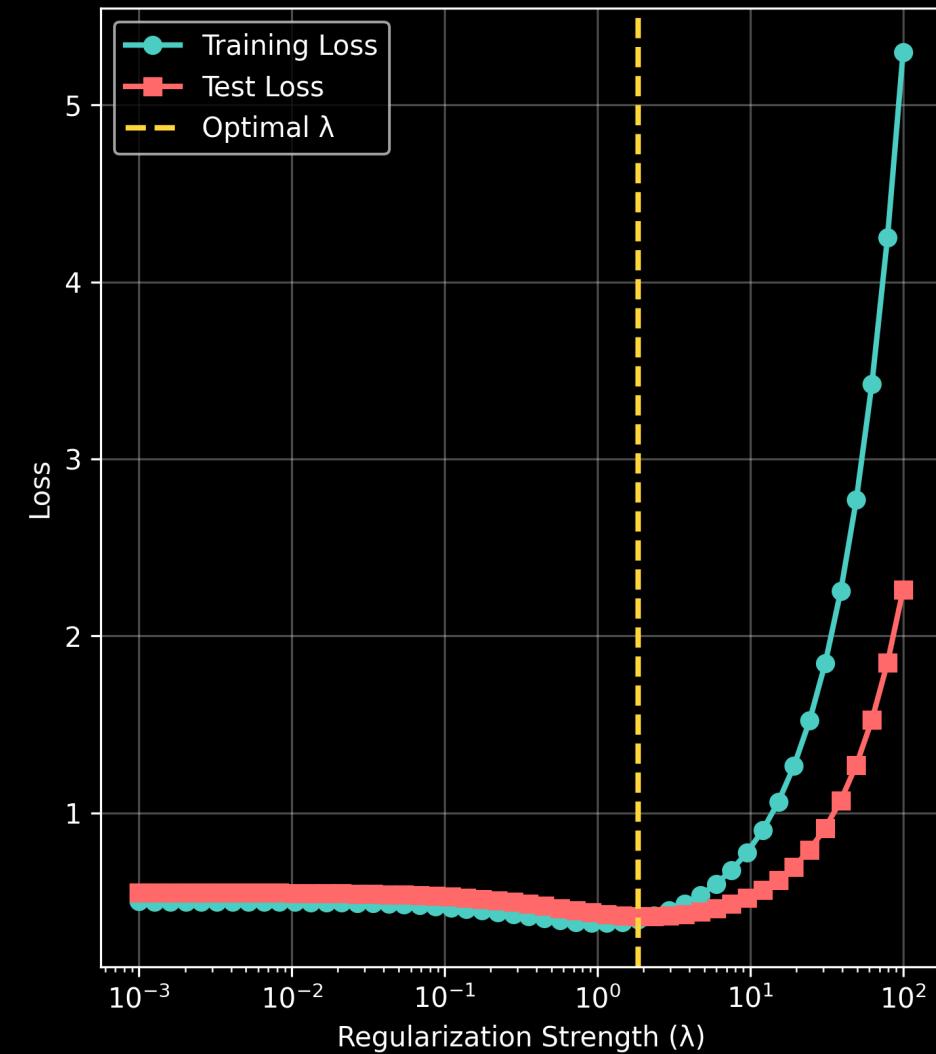


# Effect of Regularization Parameter $\lambda$

Coefficient Shrinkage with Regularization



Model Loss vs Regularization



# Regularization Parameter Values

## Common Parameter Notation:

- **$\lambda$  (lambda):** Direct regularization strength
- **C:** Inverse regularization ( $C = 1/\lambda$ )

## Effect on Student Classification:

- **Large  $\lambda$  (small C):** Simple model, may underfit
- **Small  $\lambda$  (large C):** Complex model, may overfit
- **Optimal  $\lambda$ :** Best balance for new students

<b><math>\lambda</math> Value</b>	<b>C Value</b>	<b>Model Behavior</b>
0.001	1000	Very complex
0.1	10	Moderate
1.0	1	Balanced
10	0.1	Simple
100	0.01	Very simple



# Cross-Validation for Parameter Selection

Finding Optimal Regularization for Student Classification



# What We've Covered

In this video, we've explored:

- Overfitting problems with complex student data
- L2 regularization as a solution to overfitting
- Mathematical formulation of regularized loss function
- Effect of regularization parameter  $\lambda$  on model complexity
- Cross-validation for optimal parameter selection

