

Neural networks for time series analysis

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Day 3: overfitting, convolutional and recurrent nets

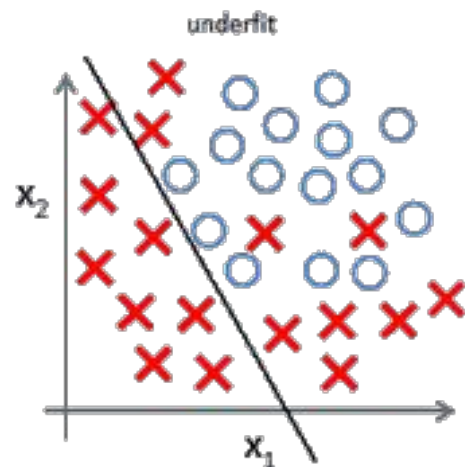
Introduction to machine
learning and time series
analysis

Data preparation and
feedforward neural
networks

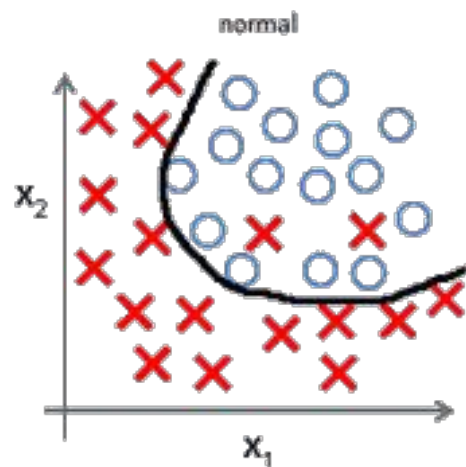
Convolutional, recurrent
neural networks and
overfitting

Building a trading
strategy and further
applications

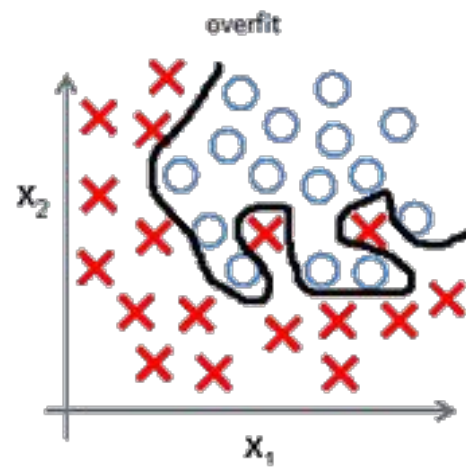
overfitting



$$g(\theta_0 + \theta_1 x_1 + \theta_2 x_2)$$



$$g(\theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1^2 + \theta_4 x_2^2 + \theta_5 x_1 x_2)$$



$$g(\theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1^2 + \theta_4 x_2^2 + \theta_5 x_1 x_2 + \theta_6 x_1^2 x_2 + \theta_7 x_1 x_2^2 + \theta_8 x_1^2 x_2^2 + \theta_9 x_1^3 + \dots)$$

data augmentation

regularization

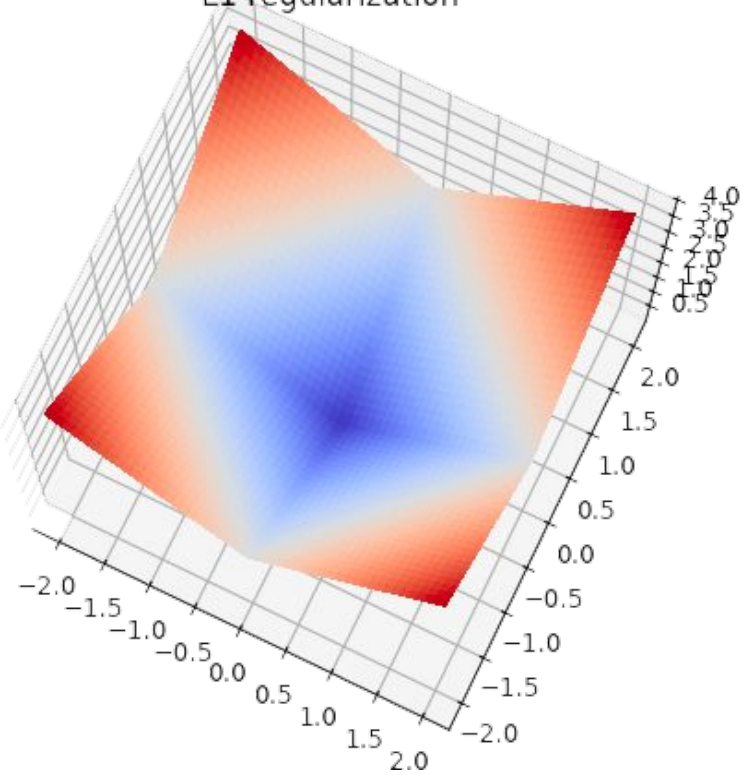
L1 Regularization

$$\text{Cost} = \sum_{i=0}^N (y_i - \sum_{j=0}^M x_{ij} W_j)^2 + \lambda \sum_{j=0}^M |W_j|$$

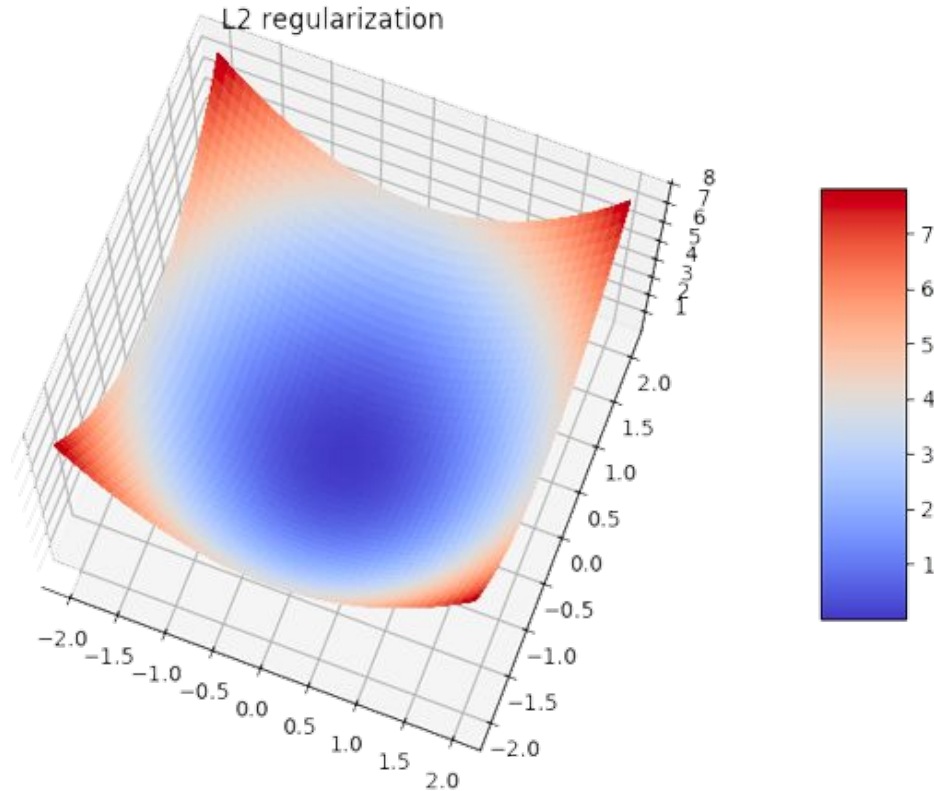
L2 Regularization

$$\text{Cost} = \underbrace{\sum_{i=0}^N (y_i - \sum_{j=0}^M x_{ij} W_j)^2}_{\text{Loss function}} + \lambda \underbrace{\sum_{j=0}^M W_j^2}_{\text{Regularization Term}}$$

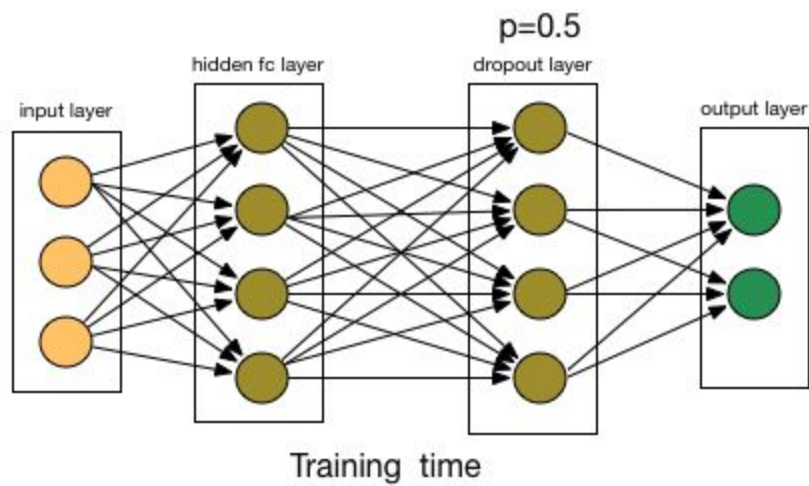
L1 regularization



L2 regularization



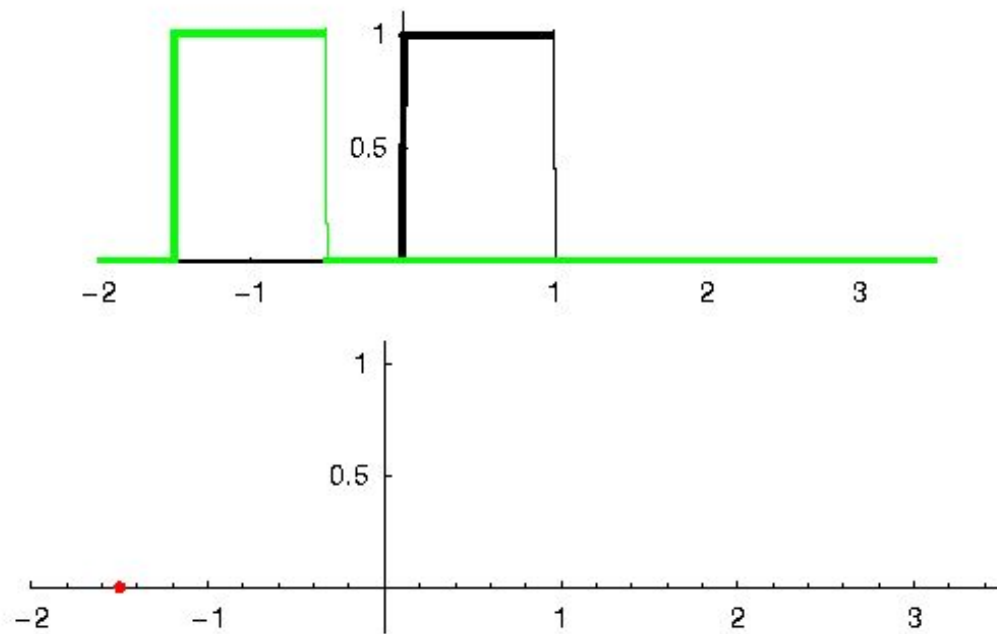
dropout



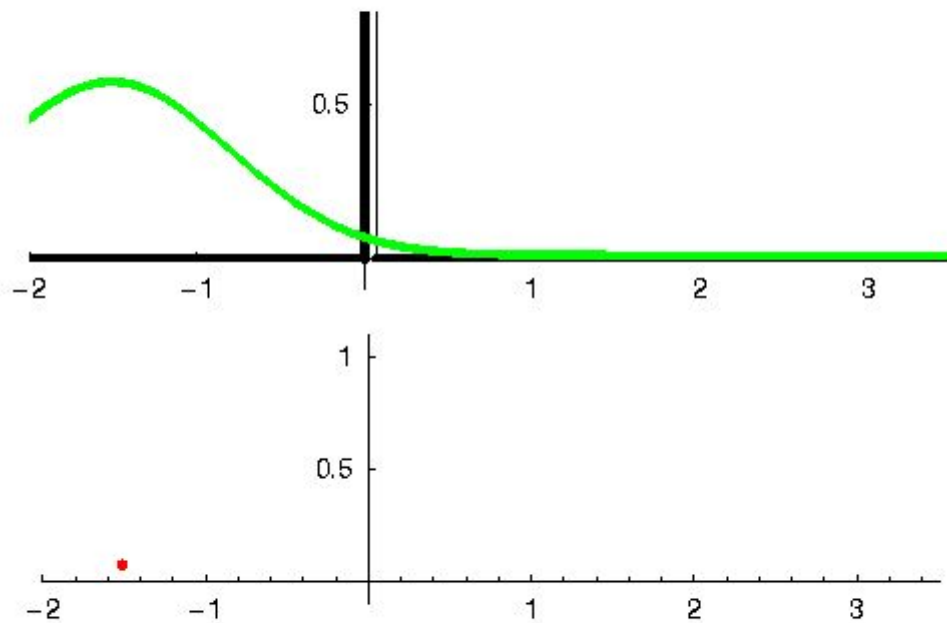
noise augmentation

convolutional neural nets

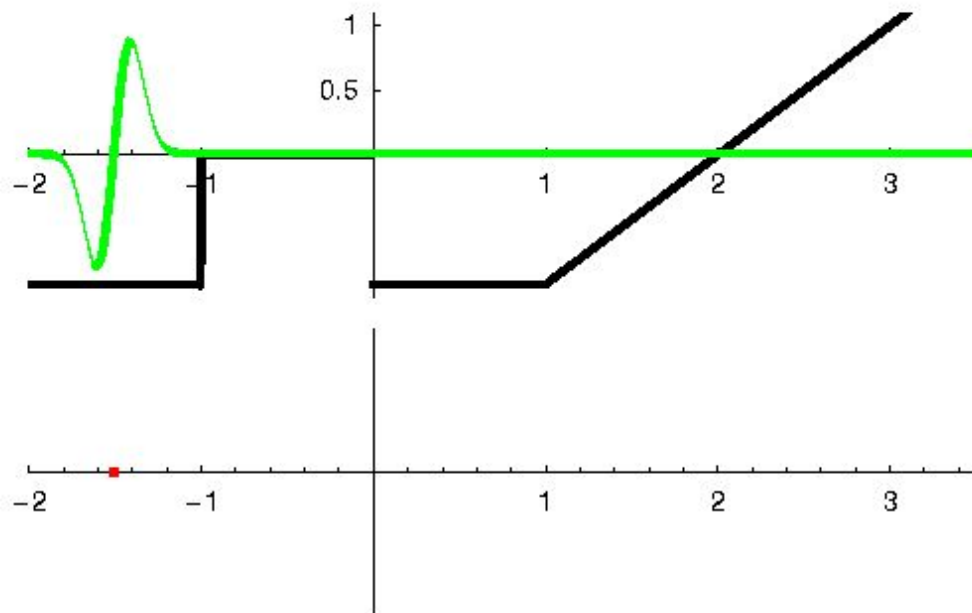
Convolution of a block with a block



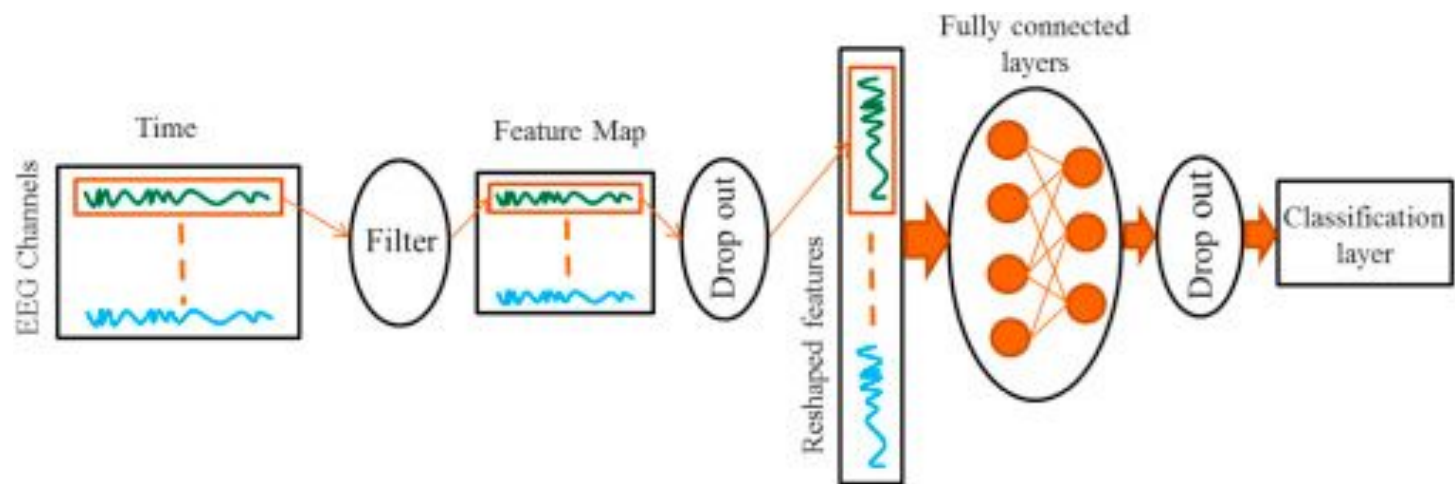
Convolution of a spike with a gaussian



edge detection



$$(f * g)(x) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(y) g(x - y) dy = \int_{-\infty}^{\infty} f(x - y) g(y) dy.$$



recurrent neural nets

b^0 is fed to next layer

