Deep Learning: Theory and Practice

Linear and Logistic Models for Classification

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Summary so far ...

- Maximum Likelihood
- Linear Least Squares Classifiers
- Logistic Regression
 - Application of ML to Logistic Regression
 - Gradient Descent
 - Coding Logistic regression





Bishop - PRML book (Chap 3)

Least Squares versus Logistic Regression





Bishop - PRML book (Chap 4)



Least Squares versus Logistic Regression





Bishop - PRML book (Chap 4)

Training, Validation and Test Set

Original Set		
Training		Testing
Training	Validation	Testing

Perceptron Algorithm

Perceptron Model [McCulloch, 1943, Rosenblatt, 1957]



Targets are binary classes [-1,1] What if the data is not linearly separable



Multi-layer Perceptron

Multi-layer Perceptron [Hopfield, 1982]

$$\mathbf{w}^{1} \cdot \mathbf{w}^{2}$$

$$\mathbf{v}^{2} = \psi \left(\mathbf{W}^{2} \phi (\mathbf{W}^{1} \mathbf{x} + \mathbf{b}^{1}) + \mathbf{b}^{2} \right)$$

$$\mathbf{w}^{1} \cdot \mathbf{b}^{2} \cdot \mathbf{b}^{2}$$

$$\phi \text{ non-linear function } (tanh, sigmoid)$$

$$\psi \text{ thresholding function}$$