

Lab 3 – Logistic Regression

We started from...

$$\min_{w \in \mathbb{R}^D} \frac{1}{n} \sum_{i=1}^n L(y_i, f_w(x_i)) + \lambda ||w||^2$$

... and picked a specific loss function ...

$$L(y_i, f_w(x_i)) = \log(1 + e^{-y_i f_w(x_i)})$$

Logistic Regression

$$\min_{w \in \mathbb{R}^D} \frac{1}{n} \sum_{i=1}^n \log(1 + e^{-y_i f_w(x_i)}) + \lambda ||w||^2$$

Despite the name we are solving a classification problem

Solving LR with Gradient Descent

$$w_t = w_{t-1} - \gamma \left(\frac{1}{n} \sum_{i=1}^n x_i \frac{-y_i}{1 + e^{y_i x_i^T w_{t-1}}} + 2\lambda w_{t-1} \right)$$

How to compute it?

How to initialize it?

When do we stop?

Solving LR with Stochastic Gradient Descent

$$w_t = w_{t-1} - \gamma_t \left(\frac{1}{n} \sum_{i=1}^n x_i \frac{-y_i}{1 + e^{y_i x_i^T w_{t-1}}} + 2\lambda w_{t-1} \right)$$

The notion of confidence

$$\frac{1}{1 + e^{-f_w(x)}}$$
$$= \frac{1}{1 + e^{-w^T x}}$$

Your objectives today

- Implementing Logistic Regression with Gradient Descent
- More specifically you will implement
 - The function **train_logreg_gd** (**Xtr**, **Ytr**, **reg_par**) to estimate the **w** using gradient descent
 - The function **predict_logreg** (**w**, **Xts**) to evaluate the function on a set of points (obtaining the prediction)
- Reason on the confidence...
- Extend to Stochastic Gradient Descent
- **IMPORTANT NOTE: we are not considering CV in this lab as we are focusing on the analysis... but we would need it for selecting the most appropriate regularization parameter!!!**

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