

Regularization (a few notes)

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Learning as search

- Initial model
 - random MLP
 - one node Decision Tree
- **Improve the model** with respect to data
 - reduce the **loss**
 - $Loss = f(Y, \hat{Y})$
 - the main part of loss is the **error**

Reduce search

- Search space can be large
- Simpler models are more likely than more complex models
 - Ockam's razor
- **Overfitting**
 - Simpler models are less prone to overfitting

Avoid complex models

- Many techniques to keep models simple
 - strong assumptions
 - linear regression
 - **penalize** more complex models
 - **complexity** parameter in Decision Trees

Regularization

- Parameter **shrinking**

- add a **term** to the Loss function
- this term penalizes **large** parameter values θ
- it is called a **regularization** term

$$Loss(Y, \hat{Y}, \theta) = Error(Y, \hat{Y}) + \lambda.f(\|\theta\|)$$

- λ is the regularization **hyper-parameter**
 - if $\lambda = 0$ there is no regularization

Regularization

- Neural Networks
 - learning is done by **minimizing** the error
 - **Backpropagation** algorithm
 - usually admits a regularization term
 - to keep $\|W\|$ small
 - MLP for regression (α is the reg. hyper parameter)

$$Loss(y, \hat{y}, \theta) = \frac{1}{2} \|y - \hat{y}\|_2^2 + \frac{\alpha}{2} \|W\|_2^2$$

Regularization

- SVM
 - SVC implementation has two regularization hyper-parameters
 - C and shrinking
 - The function to **minimize** is

$$C \sum_i \text{Loss}(y_i, \hat{y}_i) + \Omega(w)$$

- Notes
 - the i loss corresponds to ξ_i
 - the w are the parameters of the separating hyper-plane

Hyperparameters C and Ω

in SVC

- C
 - is 1 by default
 - is cost factor
 - large C penalizes errors and keeps margins small
 - uses **fewer** support vectors
 - small C allows wider margins
- shrinking
 - True by default
 - defines if a **shrinking heuristic** is used
 - the algorithm decides if w are shrunked or not
 - **may** be useful to speed up computation

References

- Books
 - Han, Kamber & Pei, Data Mining Concepts and Techniques, Morgan Kaufman.
 - Jake VanderPlas, Data Science Handbook, O'Reilly
 - Tibshirani et al., Elements of Statistical Learning
- Scikit docs
 - https://scikit-learn.org/stable/modules/neural_networks_supervised.html#regularization
 - <https://scikit-learn.org/stable/modules/svm.html#shrinking-svm>