

Regularization

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Agenda

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Machine Learning

“..gives computers the ability to learn without being explicitly programmed” (Arthur Samuel, 1959).

- Self-driving Google Car.
- Filtering for email spam.
- Netflix movie suggestions.

Machine Learning



what society thinks I do



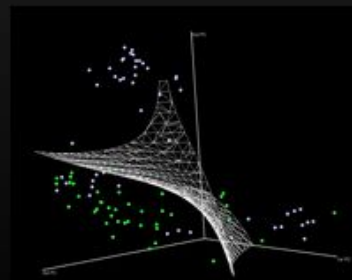
what my friends think I do



what my parents think I do

$$\begin{aligned} L_p &= \frac{1}{2} \|\mathbf{w}\|^2 - \sum_i \alpha_i y_i (\mathbf{x}_i \cdot \mathbf{w} + b) + \sum_i \alpha_i \\ \alpha_i &\geq 0, \forall i \\ \mathbf{w} &= \sum_i \alpha_i y_i \mathbf{x}_i, \sum_i \alpha_i y_i = 0 \\ \nabla \hat{g}(\theta_t) &= \frac{1}{n} \sum_{i=1}^n \nabla \ell(x_i, y_i; \theta_t) + \nabla r(\theta_t) \\ \theta_{t+1} &= \theta_t - \eta_t \nabla \ell(x_{i(t)}, y_{i(t)}; \theta_t) - \eta_t \cdot \nabla r(\theta_t) \\ \mathbb{E}_{i(t)}[\ell(x_{i(t)}, y_{i(t)}; \theta_t)] &= \frac{1}{n} \sum_i \ell(x_i, y_i; \theta_t). \end{aligned}$$

what other programmers think I do



what I think I do

```
>>> from scipy import svm
```

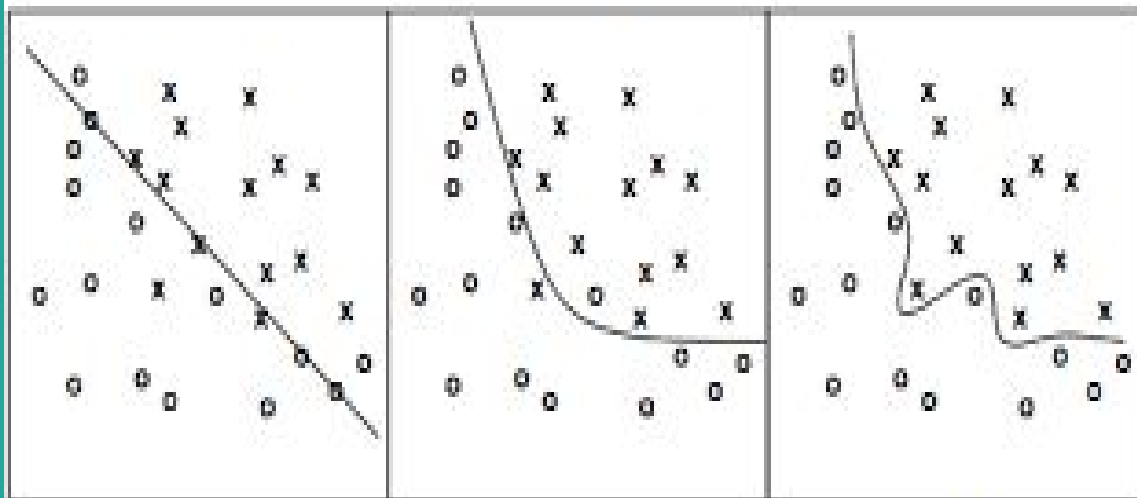
what I really do

Definition of Regularization

- It is a technique that counters the problem of overfitting by introducing tuning parameters that ease the full power of the features/variables in your dataset.
- Overfitting occurs when your predictive model aligns itself too closely with your data and not the real world.
- Two of the most used tuning parameters are:
 - L1 - Lasso
 - L2 - Ridge
- Other parameters:
 - Least Angle Regression (LARS) -
 - Elastic Net - regularized regression methods that linearly combines the penalties of the Lasso and Ridge tuning parameters. (https://en.wikipedia.org/wiki/Elastic_net_regularization)

Overfitting

“The most likely hypothesis is the **simplest** one consistent with the data.”



inadequate

good compromise

over-fitting

Lasso Regression (L1)

- This method shrinks the value of the coefficients (toward zero) using the sum of the absolute value of the coefficients.
- Lasso is preferred if we believe many features are irrelevant or we have a sparse model.

Ridge Regression (L2)

- We are again imposing a penalty on the coefficient size.
- We are shrinking the squares of the coefficients instead of using the sum of the absolute value-which was done in Lasso (L1).

Examples

Predicting number of bikes rented every hour.

DC Dept. of Transportation

- FeelsLikeTemp
- Humidity
- Windspeed
- Non-registered renters
- Registered Renters
- Number of total rentals
- DateTime
- Season
- Holiday
- WorkingDay
- Weather

Predicting Salary

- Age
- Sex
- Location
- Experience
- Race
- Family Status
- Education



Resources

- Analytics Vidhya
 - <https://www.analyticsvidhya.com/blog/2016/01/complete-tutorial-ridge-lasso-regression-python/>
- Stats website
 - <http://www.statisticshowto.com/regularization/>
- Machine Learning through Coursera/Stanford taught by Andrew Ng
 - <https://www.coursera.org/learn/machine-learning>

Conclusion and Questions

"The goal is to turn data into information, and information into insight." – Carly Fiorina, former chief executive of Hewlett-Packard Company
