-期末報告-ElasticNet

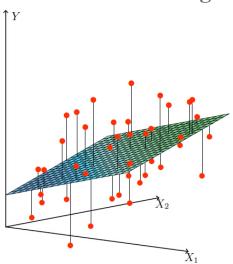
R06525055 工科碩一 吳伯彥 Regression(3)

Outline

- Algorithm Introduction
- Code Review
- Model Preview
- Live Demo
- Conclusion
- Reference

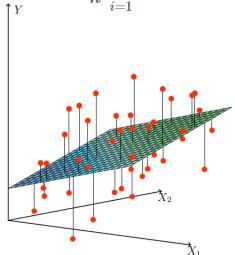
Linear Regression

Linear Regression



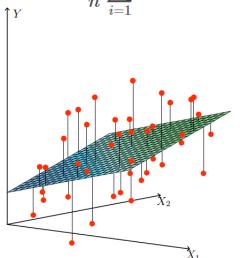
Linear Regression

Loss function:
$$\frac{1}{n}\sum_{i=1}^n(y_i-f(x_i))^2$$



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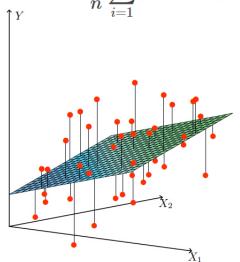


Ridge Regression

Linear Regression + L2 normalization

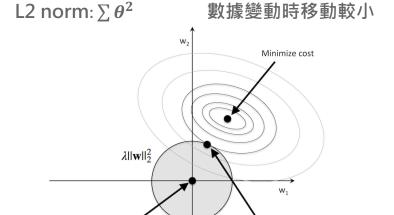
Linear Regression

Loss function: $\frac{1}{n}\sum_{i=1}^{n}(y_i-f(x_i))^2$



Ridge Regression

Linear Regression + L2 normalization



Minimize penalty

Minimize cost + penalty

Lasso Regression

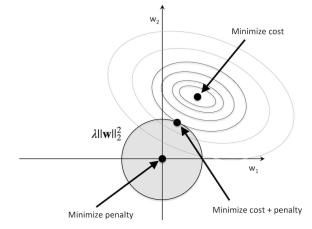
Linear Regression + L1 normalization

L1 norm: $\sum |\theta|$ 具有篩選feature的特性

Ridge Regression

Linear Regression + L2 normalization

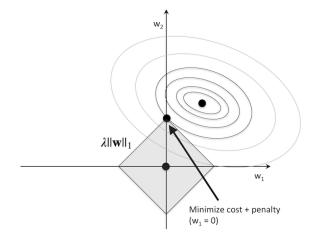
L2 norm: $\sum \theta^2$ 數據變動時移動較小



Lasso Regression

Linear Regression + L1 normalization

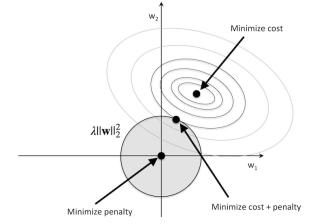
L1 norm: $\sum |\theta|$ 具有篩選feature的特性



Ridge Regression

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L2 norm: $\sum \theta^2$ 數據變動時移動較小



Elastic Net

Elastic Net

Linear Regression + L1 norm+L2 norm

L1 norm + L2 norm: $\rho \sum |\theta| + (1 - \rho) \sum \theta^2$

Elastic Net

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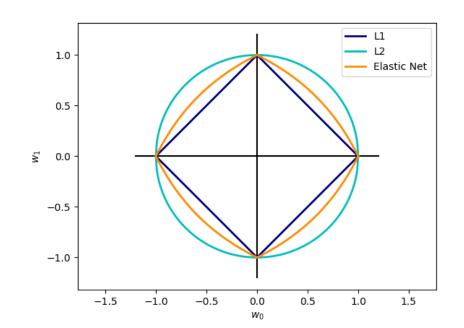
結合L1與L2的特性,Model更彈性

Linear Regression + L1 norm+L2 norm

L1 norm + L2 norm: $\rho \sum |\theta| + (1 - \rho) \sum \theta^2$

結合L1與L2的特性, Model更彈性

Elastic Net



Code Review

Param Definition(繼承algo_component中的ParamsDefinitionSet)

Code Review

利用傳入的參數建立Model並回傳(繼承algo_component中的InnalysisAlgo)

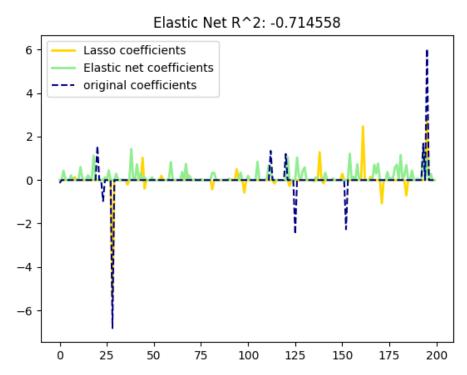
```
class ElasticNet(alc.InanalysisAlgo):
 def init (self):
    self.input_params_definition = ParamsDefinitionSet()
 def get input params definition(self):
    return self.input params definition.get params definition json list()
def do algo(self, input):
    control_params = input.algo_control.control_params
    if not self.check input_params(self.get_input_params_definition(), control_params):
        log.error("Check input params type error.")
    mode = input.algo control.mode
    data = input.algo data.data
    label = input.algo data.label
    if mode == 'training':
            model = linear model.ElasticNet(
                alpha=control_params["alpha"],
                11 ratio=control params["11 ratio"].
                fit_intercept=control_params["fit_intercept"],
                normalize=control_params["normalize"],
                precompute=control params["precompute"],
                max iter=control params["max iter"],
                copy X=control params["copy X"],
                tol=control params["tol"],
                warm start=control params["warm start"],
                positive=control params["positive"],
                random_state=control_params["random_state"],
                selection=control_params["selection"]
            model.fit(X-data, v-label)
             algo output = alc.AlgoParam(algo control={'mode': 'training', 'control params': ''},
                                             algo data={'data': data, 'label': label},
                                            algo model={'model params': model.get params(), 'model instance': model})
        except Exception as e:
             log.error(str(e))
             algo output = None
    else:
        algo_output = None
    return algo_output
```

Model Preview

產生50筆200個feature的data

分別作圖原始方程式、Lasso與Elastic net的參數值

可以看見Elastic Net比起Lasso更加圓滑一些



Live Demo

Conclusion

- 熟悉scikit learn機器學習套件的使用
- 更深入理解演算法並比較之間差異
- 學習視覺化呈現

Reference

- Scikit-learn
- STATISTICAL LEARNING AND DATA MINING III
- Ridge & LASSO & Elastic Net