

Machine Learning with Scikit-Learn

Andreas Mueller (NYU Center for Data Science, scikit-learn)

http://bit.ly/sklodsc

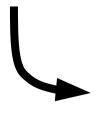
Me













Classification Regression Clustering Semi-Supervised Learning **Feature Selection Feature Extraction** Manifold Learning **Dimensionality Reduction Kernel Approximation** Hyperparameter Optimization **Evaluation Metrics** Out-of-core learning





agramfort

Alexandre Gramfort



AlexanderFabisch Alexander Fabisch



alextp Alexandre Passos



amueller Andreas Mueller



arjoly Arnaud Joly



bdholt1 Brian Holt

GaelVaroquaux

Gael Varoquaux

glouppe

jakevdp

Gilles Louppe

Jake Vanderplas

jaquesgrobler

Jaques Grobler



bthirion bthirion



chrisfilo Chris Filo Gorgole...



cournape David Cournapeau



duchesnay



Duchesnay



David Warde-Farley



fabianp Fabian Pedregosa



kuantkid Wei LI





mblondel



MechCoder







larsmans



lucidfrontier45 Shiqiao Du



Mathieu Blondel



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NelleV Varoquaux



ogrisel Olivier Grisel



paolo-losi Paolo Losi



pprett Peter Prettenhofer



robertlayton Robert Layton



ronw Ron Weiss



Satrajit Ghosh



sklearn-ci



Vlad Niculae





vmichel Vincent Michel



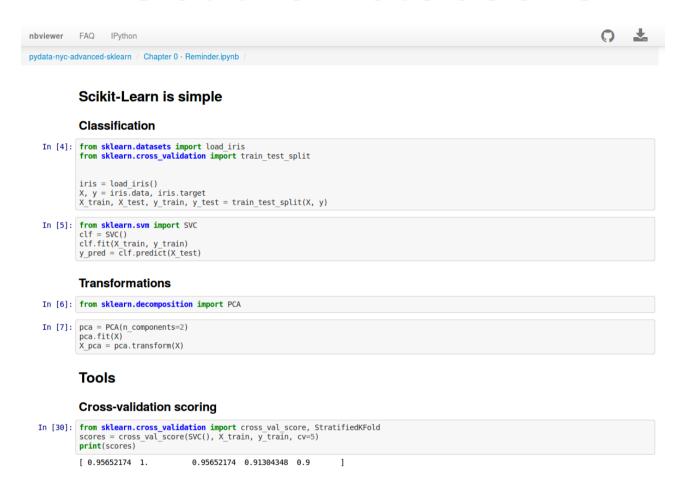
yarikoptic Yaroslav Halchenko



kastnerkyle Kyle Kastner

jnothman

Get the notebooks!



http://bit.ly/sklosdc

Documentation of scikit-learn 0.16-git

Quick Start

A very short introduction into machine learning problems and how to solve them using scikit-learn. Introduced basic concepts and conventions.

learn

User Guide

The main documentation. This contains an in-depth description of all algorithms and how to apply them.

Tutorials

Useful tutorials for developing a feel for some of scikit-learn's applications in the machine learning field.

API

The exact API of all functions and classes, as given by the docstrings. The API documents expected types and allowed features for all functions, and all parameters available for the algorithms.

Other Versions

- scikit-learn 0.15 (stable)
- scikit-learn 0.16 (development)
- scikit-learn 0.14
- scikit-learn 0.13
- scikit-learn 0.12
- Older versions

Additional Resources

Talks given, slide-sets and other information relevant to scikit-learn.

Contributing

Information on how to contribute. This also contains useful information for advanced users, for example how to build their own estimators.

Flow Chart

A graphical overview of basic areas of machine learning, and guidance which kind of algorithms to use in a given situation.

FAQ

Frequently asked questions about the project and contributing.

Hi Andy,

I just received an email from the first tutorial speaker, presenting right before you, saying he's ill and won't be able to make it.

I know you have already committed yourself to two presentations, but is there anyway you could increase your tutorial time slot, maybe just offer time to try out what you've taught? Otherwise I have to do some kind of modern dance interpretation of Python in data :-)
-Leah

Hi Andreas,

I am very interested in your Machine Learning background. I work for X Recruiting who have been engaged by Z, a worldwide leading supplier of Y. We are expanding the core engineering team and we are looking for really passionate engineers who want to create their own story and help millions of people.

Can we find a time for a call to chat for a few minutes about this?

Hi Andy,

I just received an email fraction first tutorial speaker, presenting the speaker, presenting the speaker, presenting the speaker.

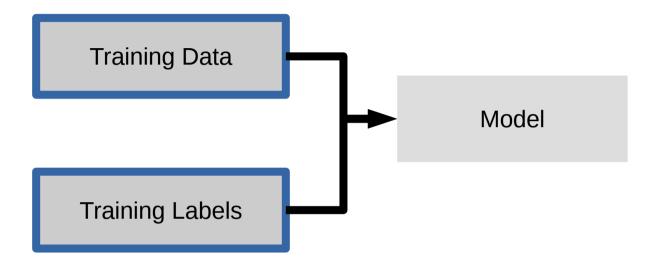
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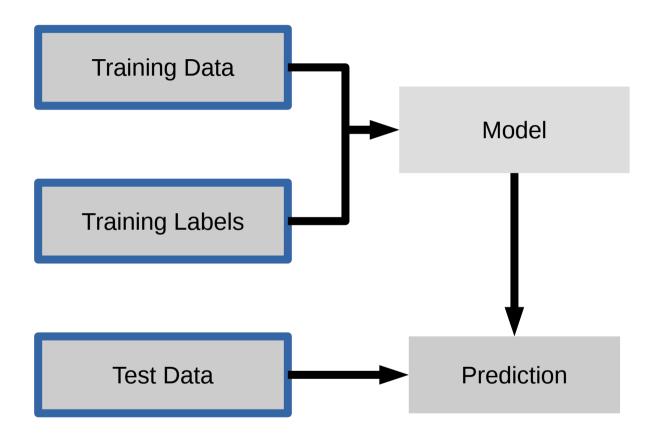
Hi Andreas,

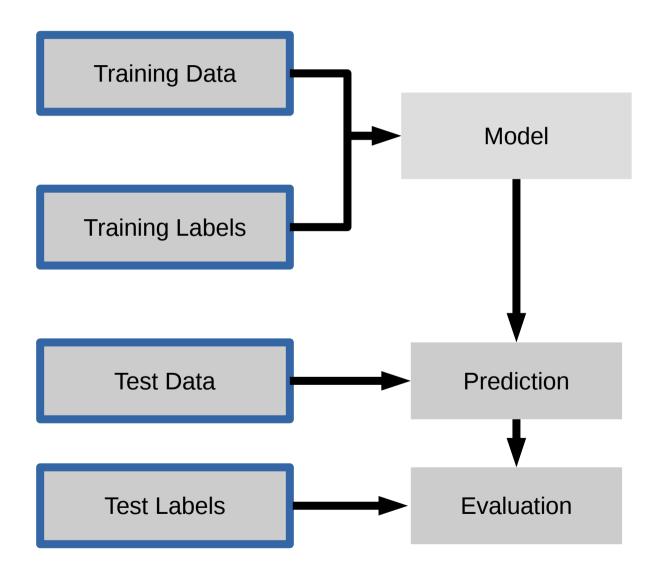
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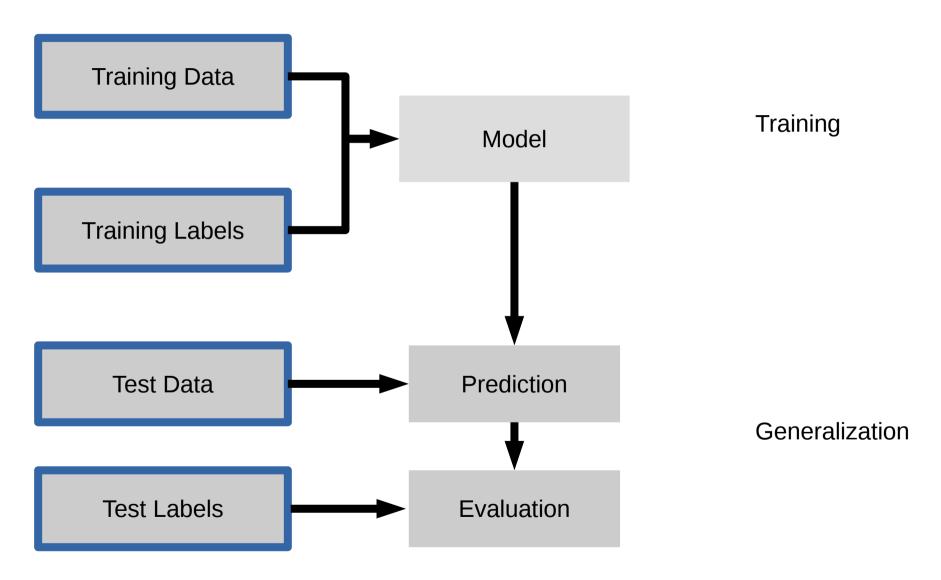
Can we find a ti minutes about t

for a few



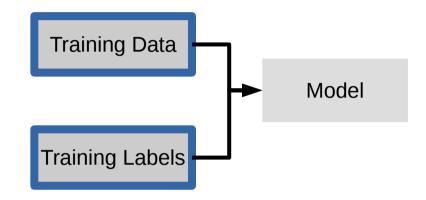






clf = RandomForestClassifier()

clf.fit(X_train, y_train)



clf = RandomForestClassifier()

clf.fit(X_train, y_train)

y_pred = clf.predict(X_test)

Training Data

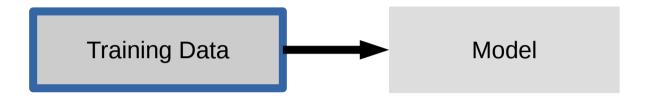
Training Data

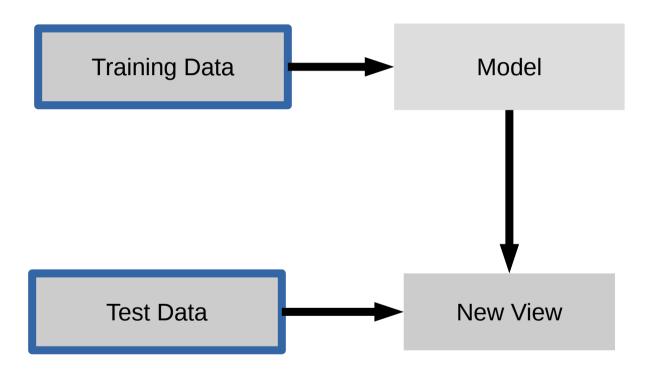
Prediction

clf = RandomForestClassifier() Training Data clf.fit(X_train, y_train) Model Training Labels y_pred = clf.predict(X_test) Test Data Prediction clf.score(X_test, y_test) Test Labels **Evaluation**

IPython Notebook: Part 1 - Introduction to Scikit-learn

Page 4-5: Cool code on how to do different algorithms at once





Unsupervised Transformations

```
pca = PCA()
pca.fit(X_train)
                                         Training Data
                                                           Model
X_new = pca.transform(X_test)
                                                        Transformation
                                          Test Data
```

IPython Notebook: Part 2 – Unsupervised Transformers

Basic API

estimator.fit(X, [y])

estimator.predict estimator.transform

Classification Preprocessing

Regression Dimensionality reduction

Clustering Feature selection

Feature extraction

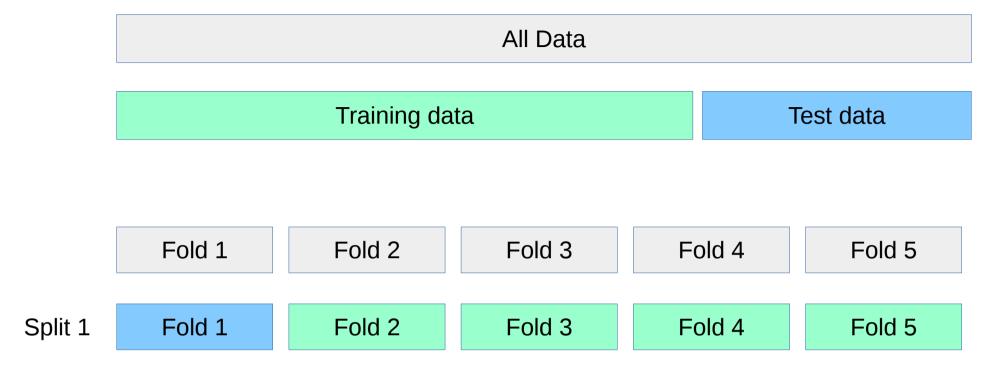
All Data			
Training data	Test data		

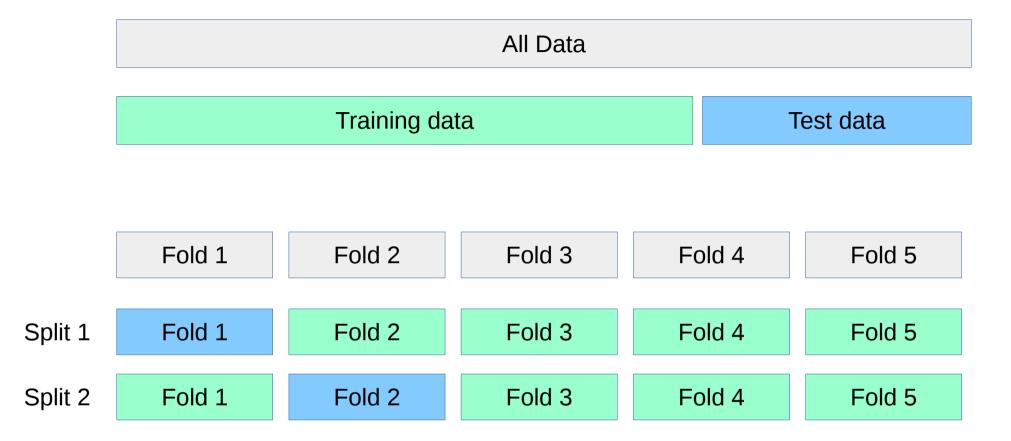
All Data

Training data

Test data

Fold 1 Fold 2 Fold 3 Fold 4 Fold 5





	All Data					
	Training data				Test data	
	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	
Split 1	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	
Split 2	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	
Split 3	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	
Split 4	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	
Split 5	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	

IPython Notebook: Part 3 - Cross-validation

```
In [2]: clf = SVC()
  clf.fit(X_train, y_train)
  y_pred = clf.predict(X_test)
```

How do we find the right parameters?

All Data			
Training data	Test data		

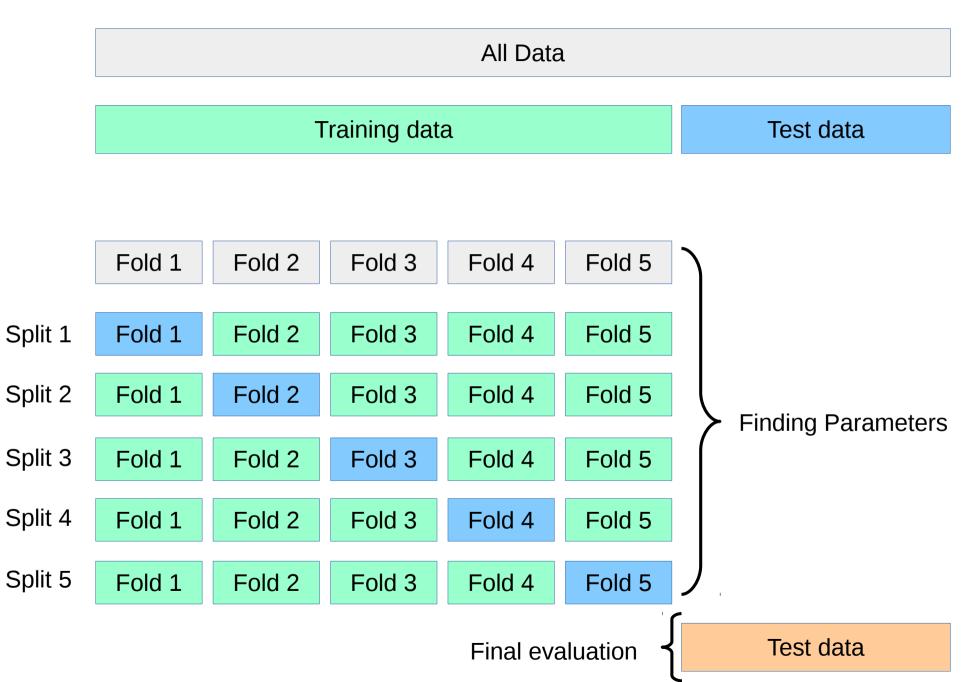
All Data

Training data

Test data

	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5
Split 1	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5
Split 2	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5
Split 3	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5
Split 4	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5
Split 5	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5

Test data



SVC(C=0.001, gamma=0.001)

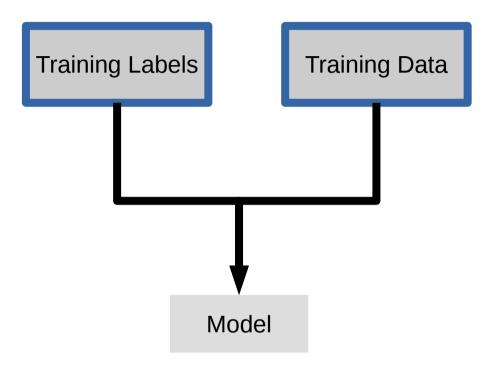
SVC(C=0.001, SVC(C=0.01, SVC(C=0.1, SVC(C=1, SVC(C=10, gamma=0.001) gamma=0.001) gamma=0.001) gamma=0.001)

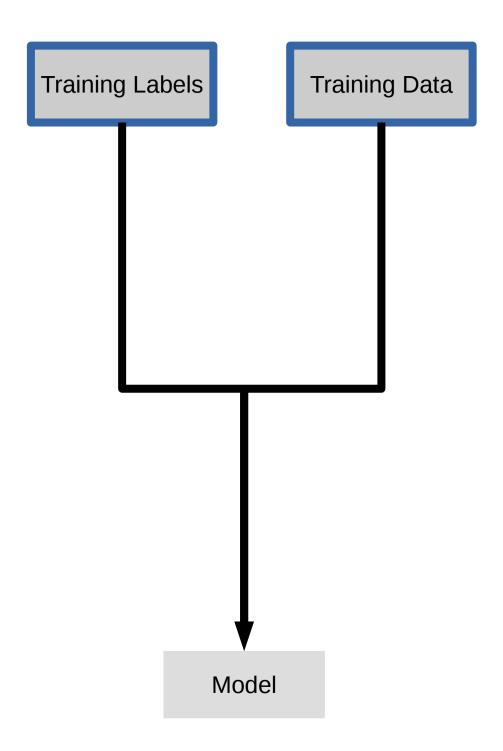
SVC(C=0.001,	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
gamma=0.001)	gamma=0.001)	gamma=0.001)	gamma=0.001)	gamma=0.001)
SVC(C=0.001, gamma=0.01)	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
	gamma=0.01)	gamma=0.01)	gamma=0.01)	gamma=0.01)

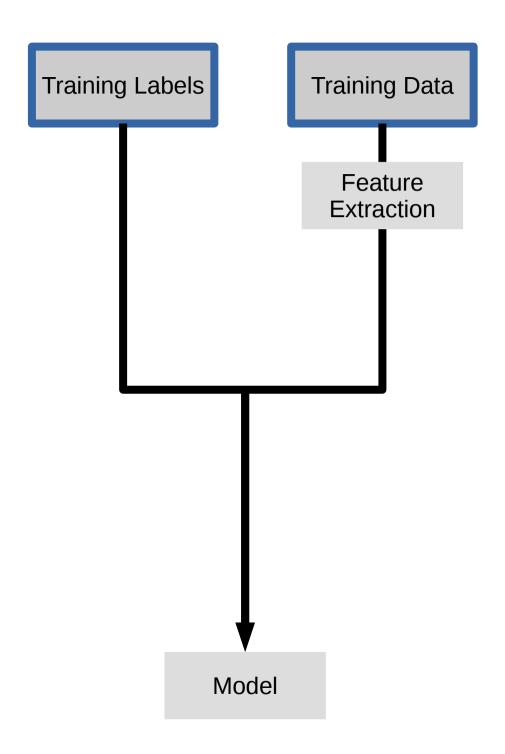
SVC(C=0.001, gamma=0.001)	SVC(C=0.01, gamma=0.001)	SVC(C=0.1, gamma=0.001)	SVC(C=1, gamma=0.001)	SVC(C=10, gamma=0.001)
SVC(C=0.001, gamma=0.01)	SVC(C=0.01, gamma=0.01)	SVC(C=0.1, gamma=0.01)	SVC(C=1, gamma=0.01)	SVC(C=10, gamma=0.01)
SVC(C=0.001, gamma=0.1)	SVC(C=0.01, gamma=0.1)	SVC(C=0.1, gamma=0.1)	SVC(C=1, gamma=0.1)	SVC(C=10, gamma=0.1)

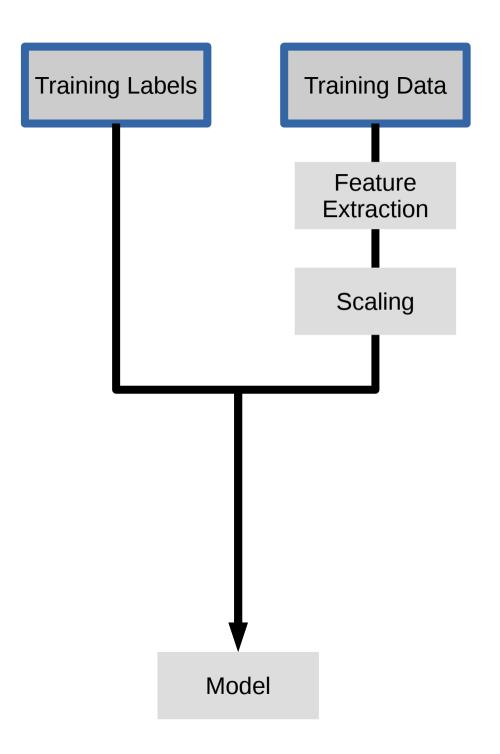
SVC(C=0.001,	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
gamma=0.001)	gamma=0.001)	gamma=0.001)	gamma=0.001)	gamma=0.001)
SVC(C=0.001,	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
gamma=0.01)	gamma=0.01)	gamma=0.01)	gamma=0.01)	gamma=0.01)
SVC(C=0.001,	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
gamma=0.1)	gamma=0.1)	gamma=0.1)	gamma=0.1)	gamma=0.1)
SVC(C=0.001, gamma=1)	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
	gamma=1)	gamma=1)	gamma=1)	gamma=1)
SVC(C=0.001,	SVC(C=0.01,	SVC(C=0.1,	SVC(C=1,	SVC(C=10,
gamma=10)	gamma=10)	gamma=10)	gamma=10)	gamma=10)

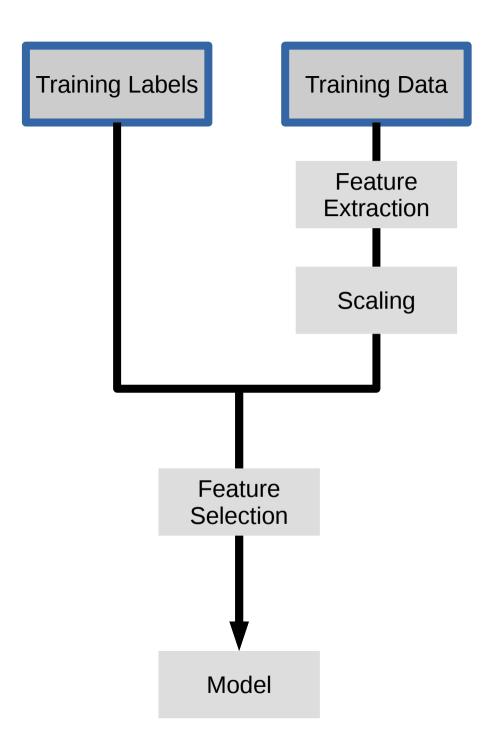
IPython Notebook: Part 4 – Grid Searches

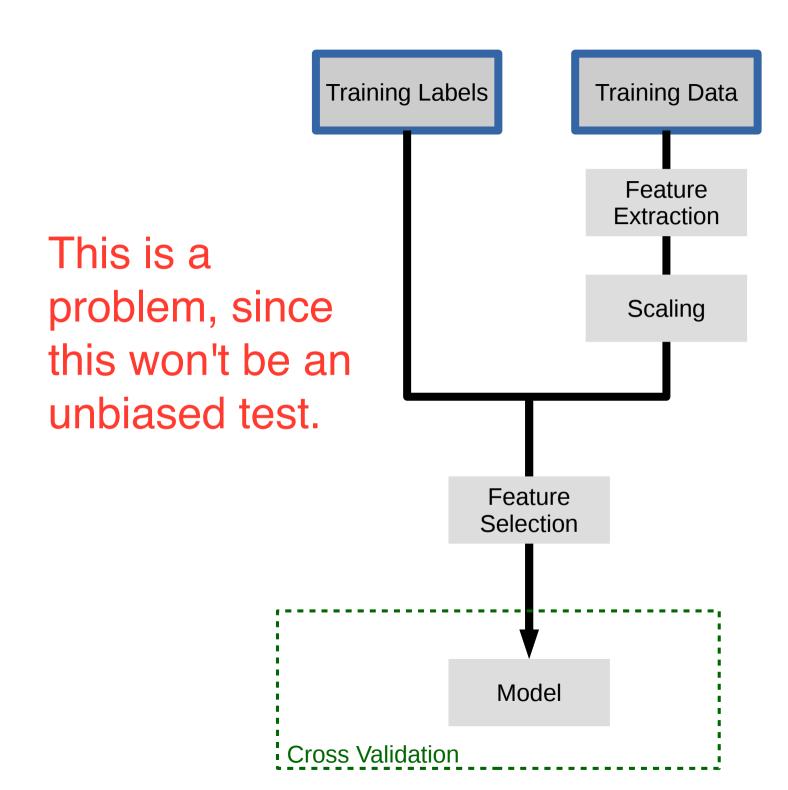


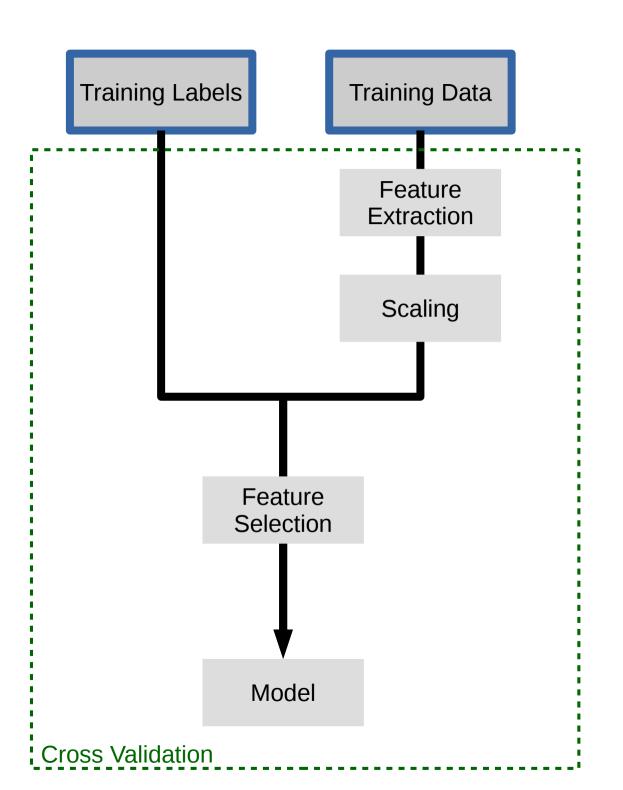












IPython Notebook: Part 5 - Preprocessing and Pipelines

Do cross-validation over all steps jointly. Keep a separate test set until the very end.

CountVectorizer / TfidfVectorizer

"This is how you get ants."

```
"This is how you get ants."

tokenizer

['this', 'is', 'how', 'you', 'get', 'ants']
```

```
"This is how you get ants."

tokenizer

['this', 'is', 'how', 'you', 'get', 'ants']

Build a vocabulary over all documents

['aardvak', 'amsterdam', 'ants', ... 'you', 'your', 'zyxst']
```

```
"This is how you get ants."
                          tokenizer
['this', 'is', 'how', 'you', 'get', 'ants']
                            Build a vocabulary over all documents
['aardvak', 'amsterdam', 'ants', ... 'you',
               'your', 'zyxst']
                          Sparse matrix encoding
  aardvak ants get you zyxst
     [0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0]
```

Application: Insult detection

Application: Insult detection

i really don't understand your point. It seems that you are mixing apples and oranges.

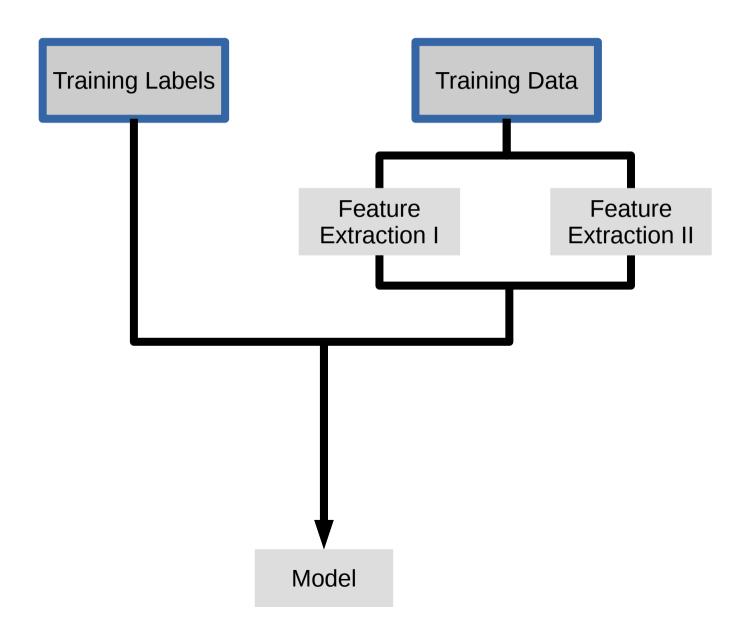
Application: Insult detection

i really don't understand your point. It seems that you are mixing apples and oranges.

Clearly you're a fucktard.

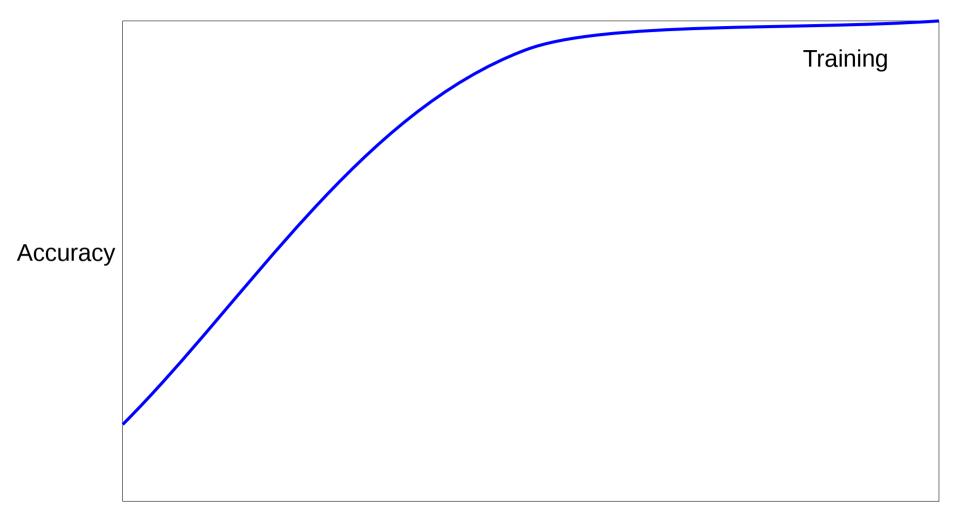
IPython Notebook: Part 6 - Working With Text Data

Feature Union



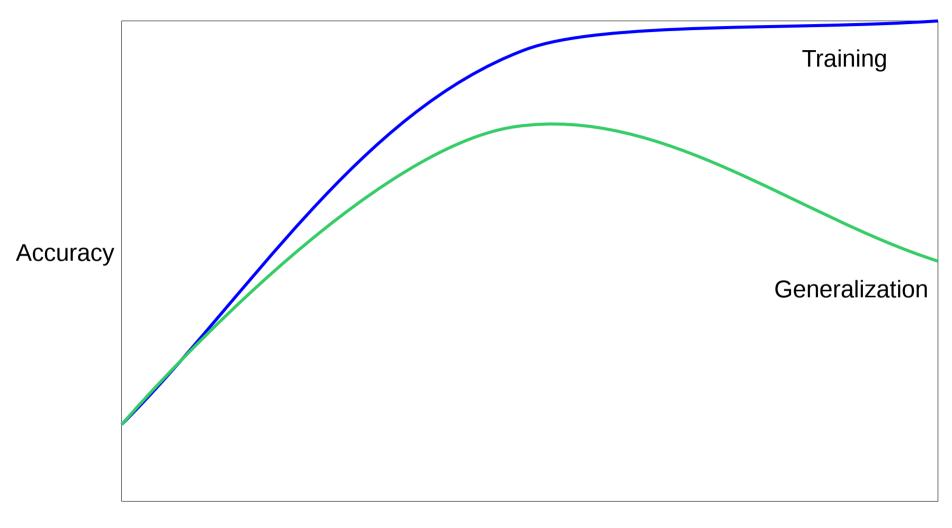
IPython Notebook: Part 7 – FeatureUnion

Overfitting and Underfitting



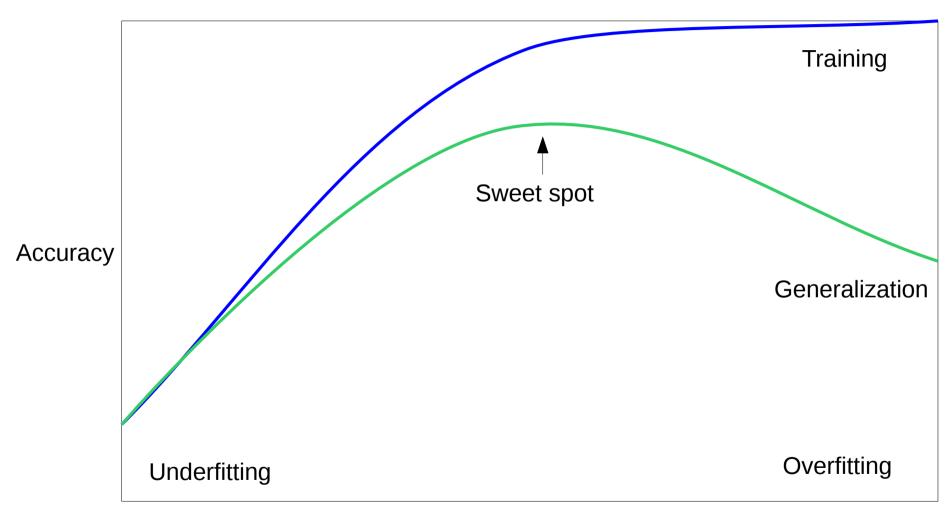
Model complexity

Overfitting and Underfitting



Model complexity

Overfitting and Underfitting



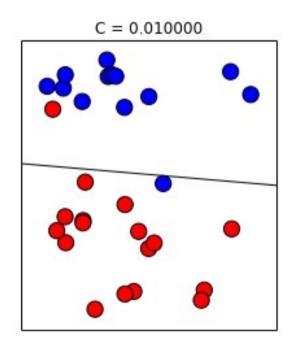
Model complexity

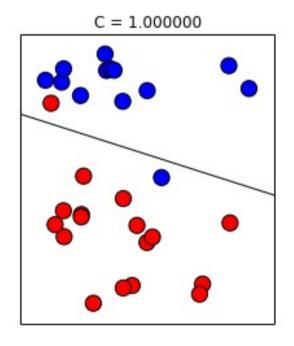
Linear SVM

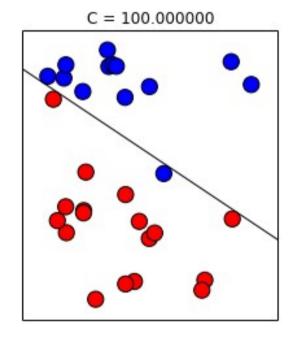
$$\hat{y} = \operatorname{sign}(w_0 + \sum_i w_i x_i)$$

Linear SVM

$$\hat{y} = \operatorname{sign}(w_0 + \sum_i w_i x_i)$$



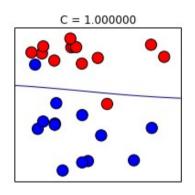


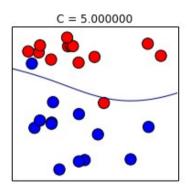


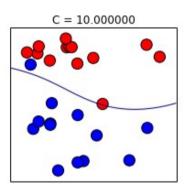
$$\hat{y} = \operatorname{sign}(\alpha_0 + \sum_j \alpha_j y_j k(\mathbf{x}^{(\mathbf{j})}, \mathbf{x}))$$

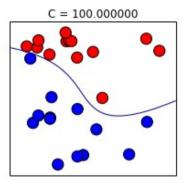
$$\hat{y} = \operatorname{sign}(\alpha_0 + \sum_j \alpha_j y_j k(\mathbf{x}^{(j)}, \mathbf{x}))$$
$$k(\mathbf{x}, \mathbf{x}') = \exp(-\gamma ||\mathbf{x} - \mathbf{x}'||^2)$$

$$\hat{y} = \operatorname{sign}(\alpha_0 + \sum_j \alpha_j y_j k(\mathbf{x}^{(j)}, \mathbf{x}))$$
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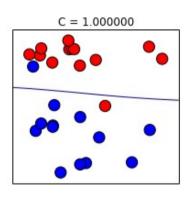


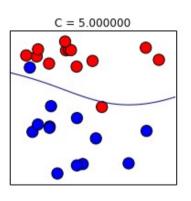


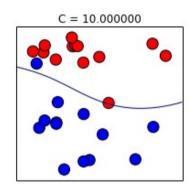


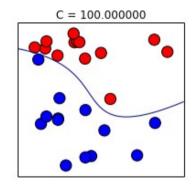


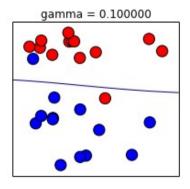
$$\hat{y} = \operatorname{sign}(\alpha_0 + \sum_j \alpha_j y_j k(\mathbf{x}^{(j)}, \mathbf{x}))$$
$$k(\mathbf{x}, \mathbf{x}') = \exp(-\gamma ||\mathbf{x} - \mathbf{x}'||^2)$$

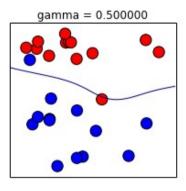


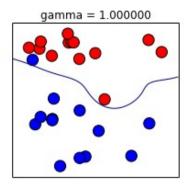


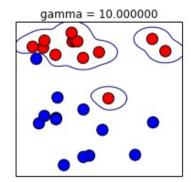


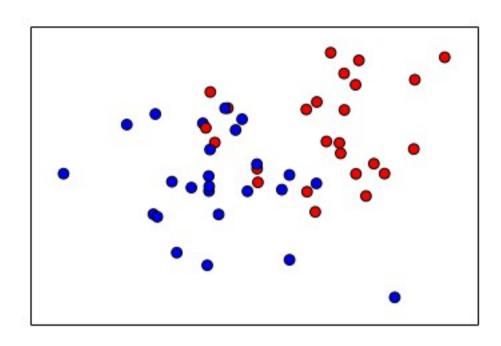


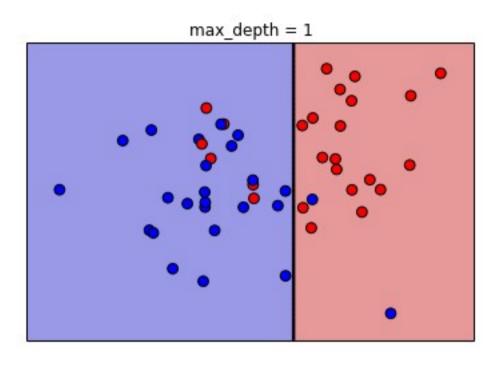


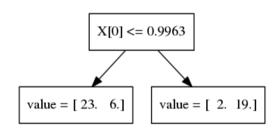


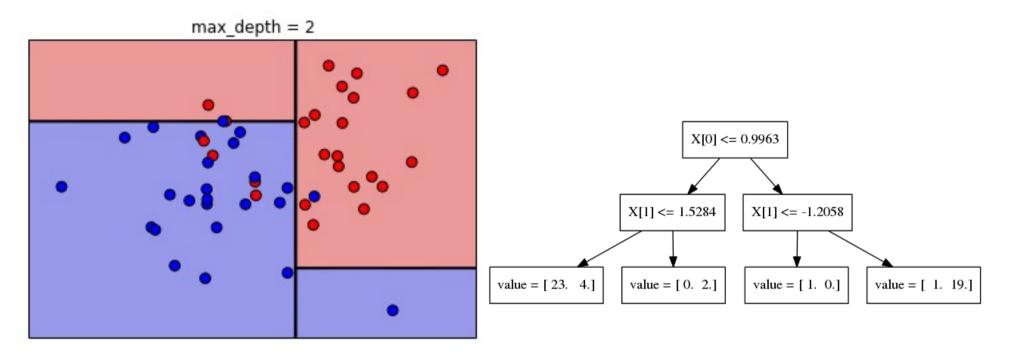


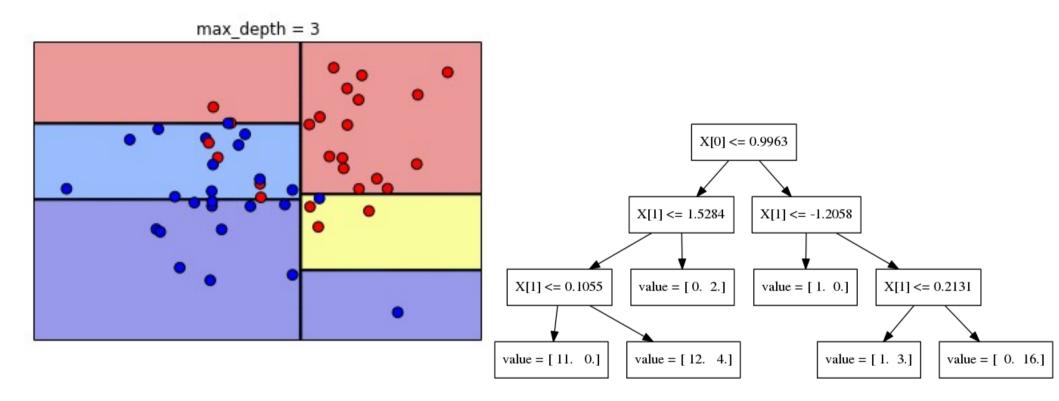


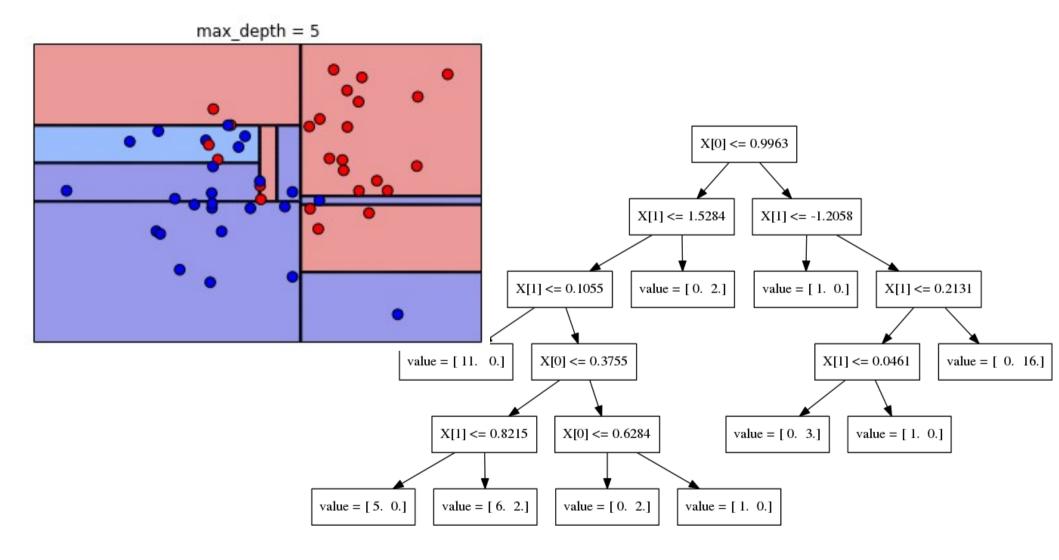




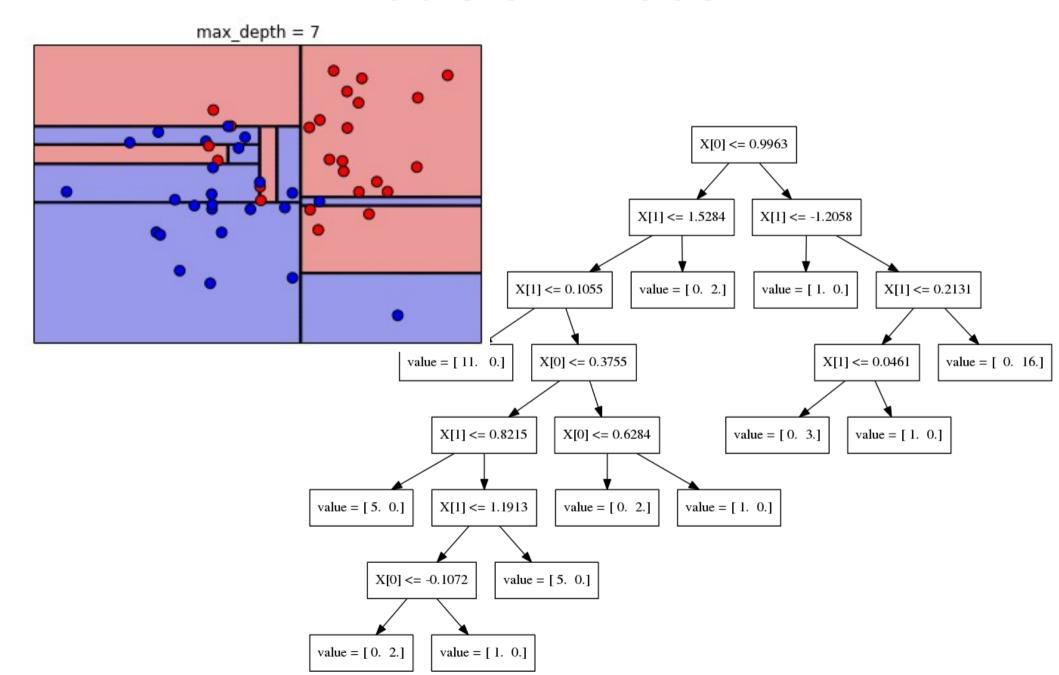




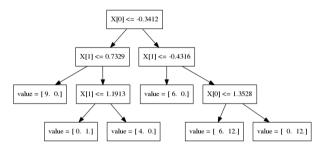




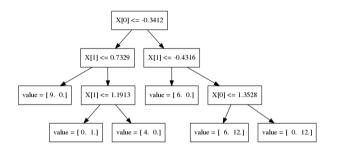
Decision Trees

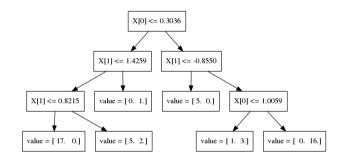


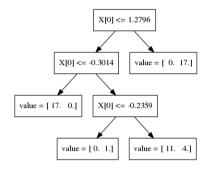
Random Forests



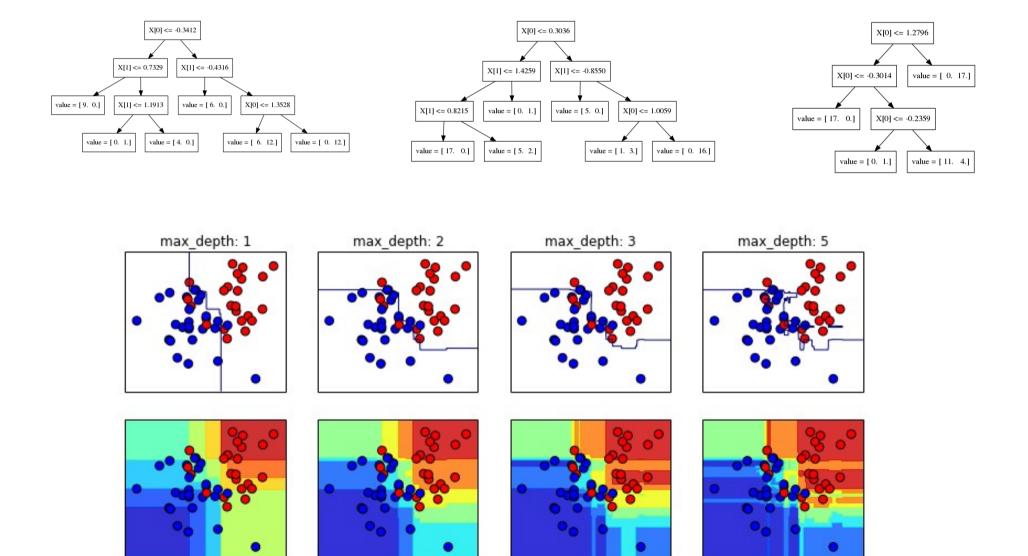
Random Forests



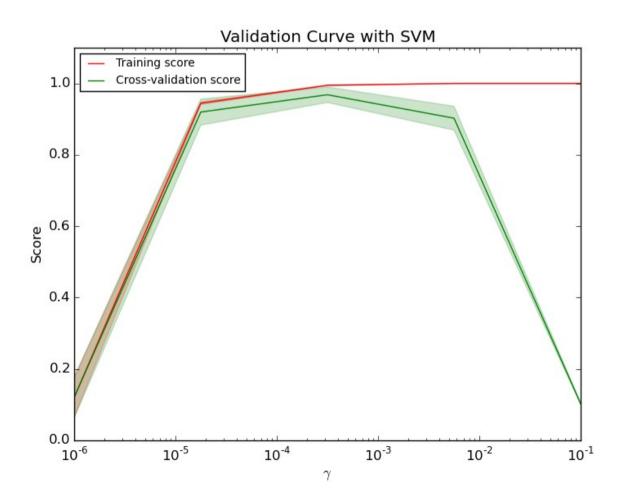




Random Forests

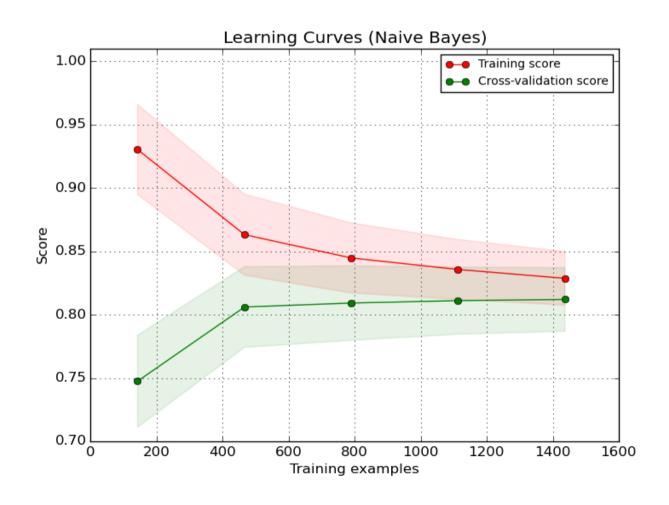


Validation Curves

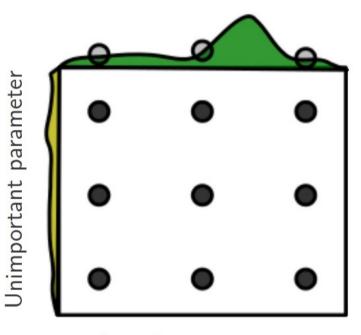


Learning Curves

train_sizes, train_scores, test_scores = learning_curve(
 estimator, X, y, train_sizes=train_sizes)

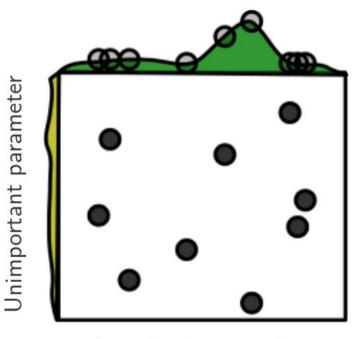


Grid Layout



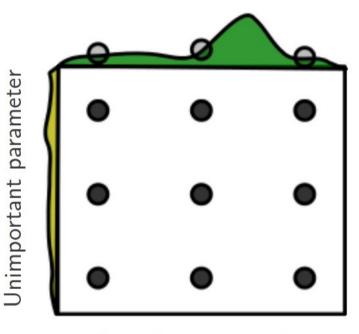
Important parameter

Random Layout



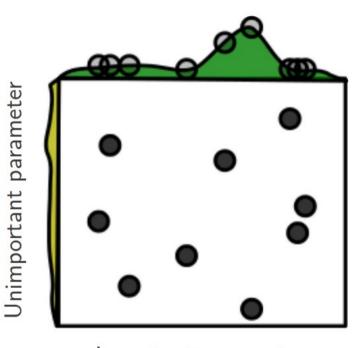
Important parameter

Grid Layout



Important parameter

Random Layout



Important parameter

Step-size free for continuous parameters Decouples runtime from search-space size Robust against irrelevant parameters

- Always use distributions for continuous variables.
- Don't use for low dimensional spaces.
- Future: Bayesian optimization based search.

IPython Notebook: Part 8 – Randomized Parameter Search

Generalized Cross-Validation and Path Algorithms

IPython Notebook: Part 9 – Generalized Cross-Validation

Linear Models	Feature Selection	Tree-Based models [possible]
LogisticRegressionCV [new]	RFECV	[DecisionTreeCV]
RidgeCV		[RandomForestClassifierCV]
RidgeClassifierCV		[GradientBoostingClassifierCV]
LarsCV		
ElasticNetCV 		

Scoring Functions

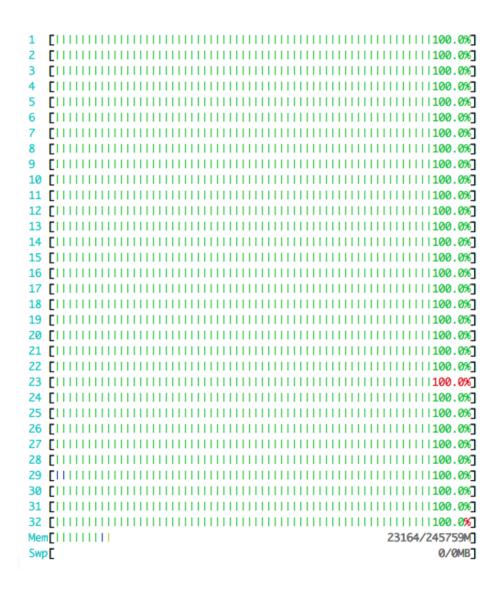
GridSeachCV RandomizedSearchCV cross_val_score ...CV

Default: Accuracy (classification) R2 (regression)

IPython Notebook: Part 10 – Scoring Metrics

Out of Core Learning

Or: save ourself the effort



Think twice!

- Old laptop: 4GB Ram
- 1073741824 float32
- Or 1mio data points with 1000 features
- EC2: 256 GB Ram
- 68719476736 float32
- Or 68mio data points with 1000 features

Supported Algorithms

- All SGDClassifier derivatives
- Naive Bayes
- MinibatchKMeans
- Birch
- IncrementalPCA
- MiniBatchDictionaryLearning

IPython Notebook: Part 11 – Out Of Core Learning

Stateless Transformers

- Normalizer
- HashingVectorizer
- RBFSampler (and other kernel approx)

Bag Of Word Representations

CountVectorizer / TfidfVectorizer

```
"This is how you get ants."
                          tokenizer
['this', 'is', 'how', 'you', 'get', 'ants']
                            Build a vocabulary over all documents
['aardvak', 'amsterdam', 'ants', ... 'you',
               'your', 'zyxst']
                          Sparse matrix encoding
  aardvak ants get you zyxst
     [0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0]
```

Hashing Trick

HashingVectorizer

```
"This is how you get ants."
                             tokenizer
   ['this', 'is', 'how', 'you', 'get', 'ants']
                               hashing
[hash('this'), hash('is'), hash('how'), hash('you'),
             hash('get'), hash('ants')]
= [832412, 223788, 366226, 81185, 835749, 173092]
                             Sparse matrix encoding
      aardvak ants
                        get
                              you zyxst
        [0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0, 1, 0, ..., 0]
```

IPython Notebook: Part 12 – Out Of Core Learning for Text

Documentation of scikit-learn 0.16-git

Quick Start

A very short introduction into machine learning problems and how to solve them using scikit-learn. Introduced basic concepts and conventions.

learn

User Guide

The main documentation. This contains an in-depth description of all algorithms and how to apply them.

Tutorials

Useful tutorials for developing a feel for some of scikit-learn's applications in the machine learning field.

API

The exact API of all functions and classes, as given by the docstrings. The API documents expected types and allowed features for all functions, and all parameters available for the algorithms.

Other Versions

- scikit-learn 0.15 (stable)
- scikit-learn 0.16 (development)
- scikit-learn 0.14
- scikit-learn 0.13
- scikit-learn 0.12
- Older versions

Additional Resources

Talks given, slide-sets and other information relevant to scikit-learn.

Contributing

Information on how to contribute. This also contains useful information for advanced users, for example how to build their own estimators.

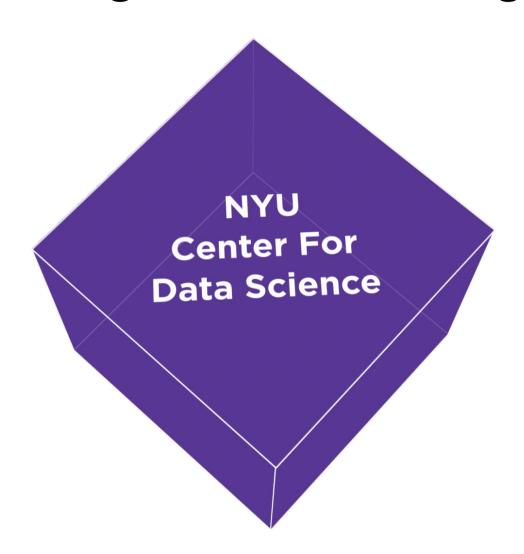
Flow Chart

A graphical overview of basic areas of machine learning, and guidance which kind of algorithms to use in a given situation.

FAQ

Frequently asked questions about the project and contributing.

CDS is hiring Research Engineers



Thank you for your attention.



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