Python第三方库

outline

- □ 数值计算 numpy
- □ 数据处理分析 pandas
- □ 可视化 matplotlib/seaborn
- □ 机器学习 Sklearn / keras
- □ 交互 pygame
- □ 网络 Selenium etc...

(今天讲红色字)

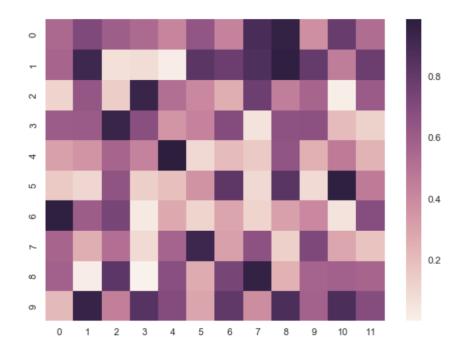
硬广告: python爬虫项目班开课了。。



seaborn

Plot a heatmap for a numpy array:

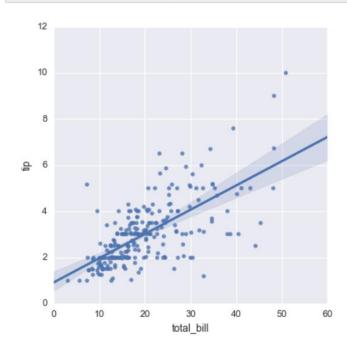
```
>>> import numpy as np; np.random.seed(0)
>>> import seaborn as sns; sns.set()
>>> uniform_data = np.random.rand(10, 12)
>>> ax = sns.heatmap(uniform_data)
```





seaborn

```
>>> import seaborn as sns; sns.set(color_codes=True)
>>> tips = sns.load_dataset("tips")
>>> g = sns.lmplot(x="total_bill", y="tip", data=tips)
```



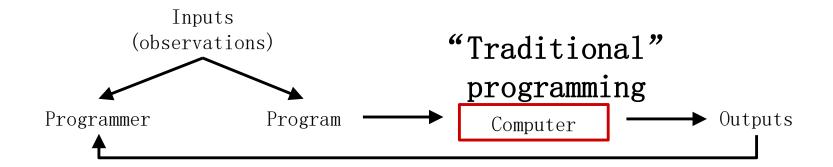


第三方库的安装

- ☐ Pip install
- ☐ Use anaconda, if possible

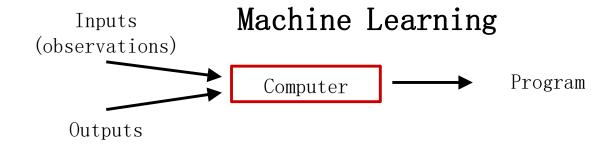


What is Machine Learning?



Machine Learning is the field of study that gives computers the ability to learn without being explicitly programmed.

-- Arthur Samuel (1959)





Examples of Machine Learning



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By Steve Jurvetson [CC BY 2.0]



3 Types of Learning

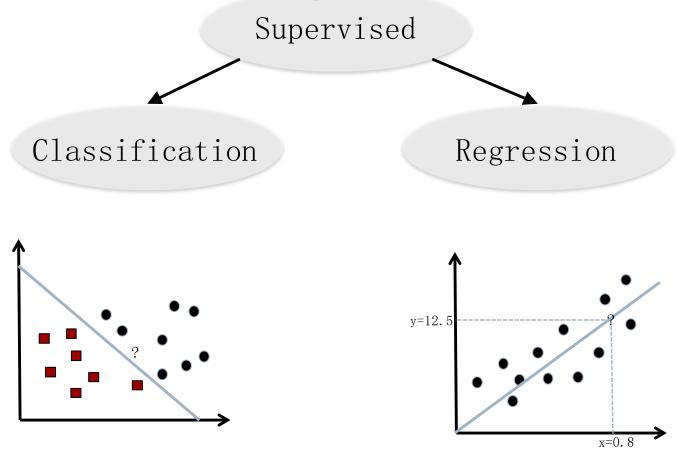
Supervised Unsupervised Reinforcement

- ➤ Discover structure in ➤ Learning by "doing" unlabeled data with delayed reward
- > E.g., Document clustering

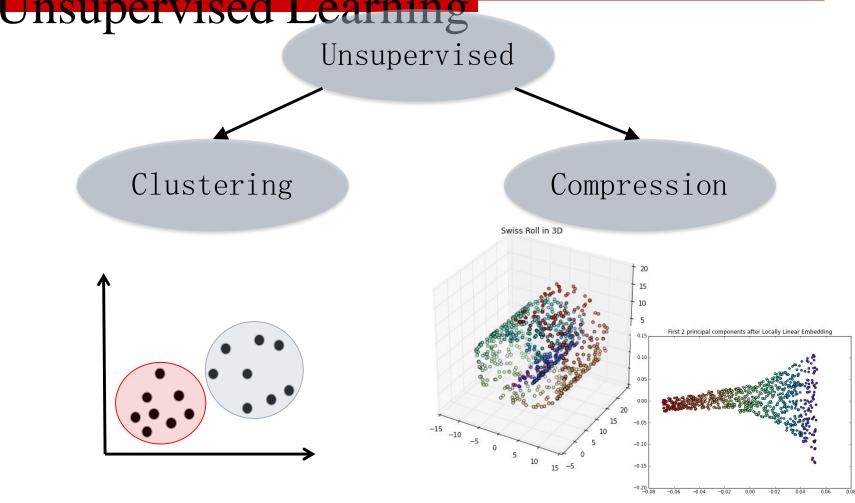
- ➤ E.g., Chess computer



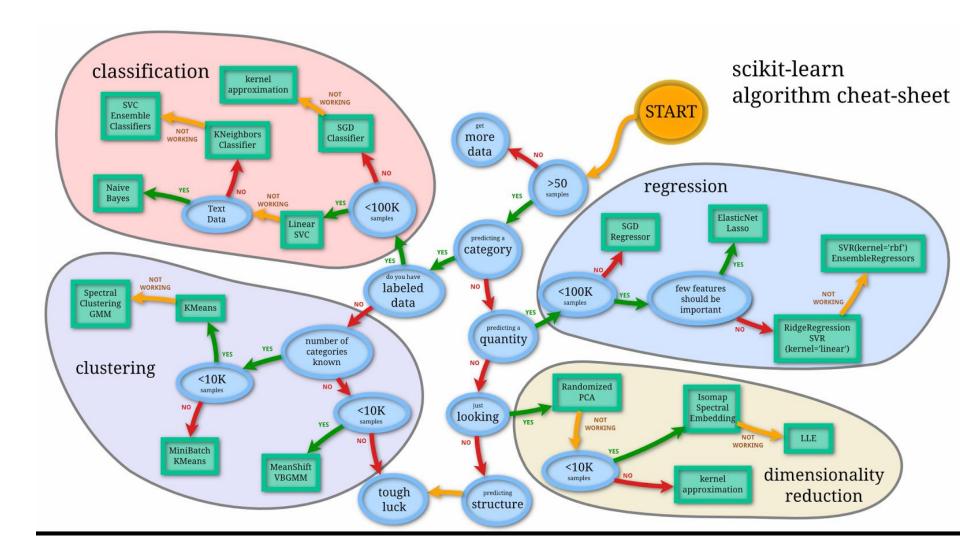
Supervised Learning









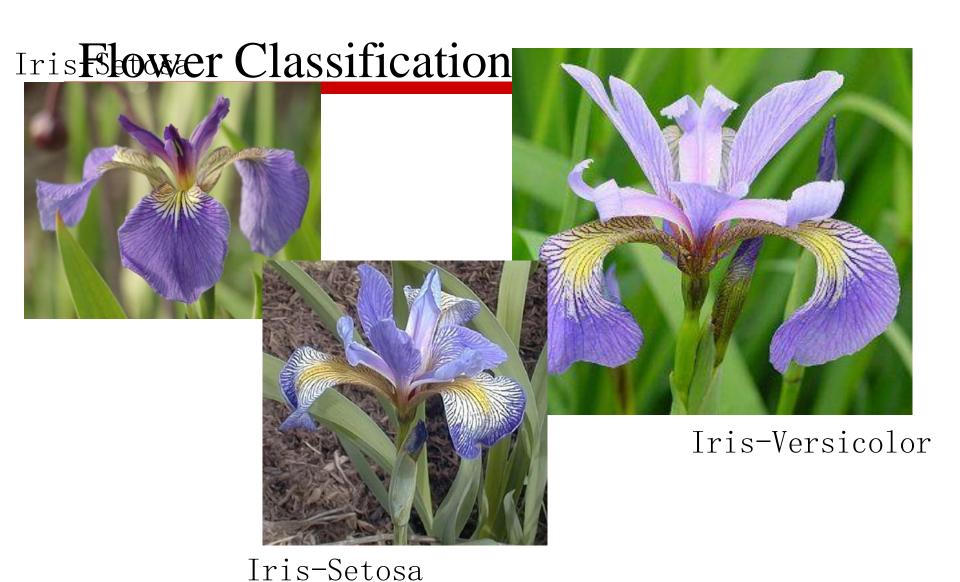




The simplest Sklearn workflow

```
train_x, train_y, test_x, test_y = getData()
model = somemodel()
model.fit(train_x,train_y)
predictions = model.predict(test x)
score = score_function(test_y, predictions)
```



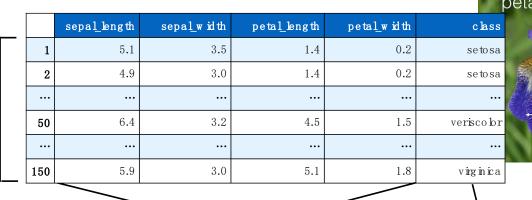




Data Representation In stances (samples, observations)

$\mathbb{R} \mathbb{S}$

https://archive.ics.uci.edu/ml/datasets/Iris



Features (attributes, dimensions)

Classes (targets)

sepal

$$\mathbf{X} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1D} \\ x_{21} & x_{22} & \cdots & x_{2D} \\ x_{31} & x_{32} & \cdots & x_{3D} \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \vdots \\ x_{N1} & x_{N2} & \cdots & x_{ND} \end{bmatrix}$$

$$\mathbf{y} = [y_1, y_2, y_3, \cdots y_N]$$



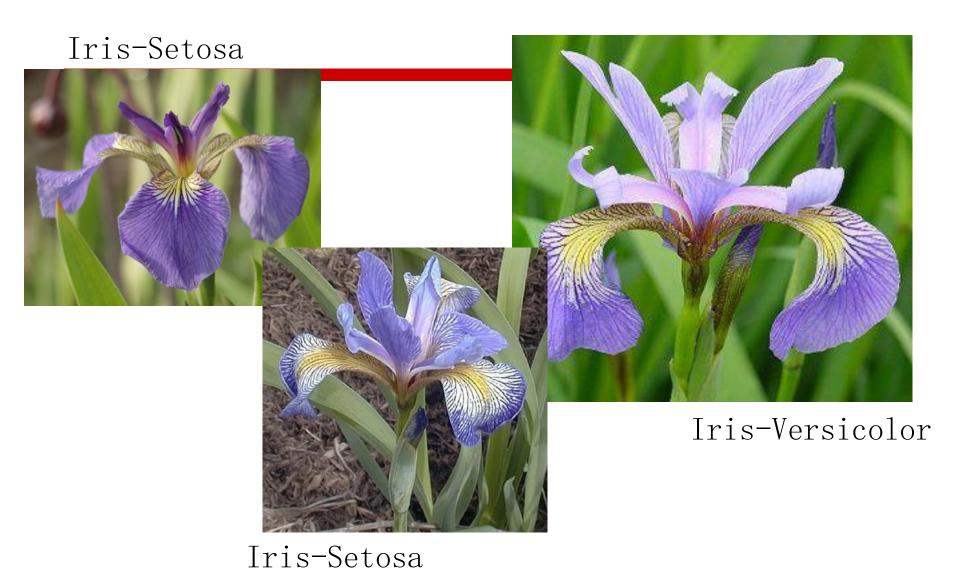
```
In [2]: from sklearn.datasets import load_iris
    iris = load_iris()

The resulting dataset is a Bunch object: you can see what's available using the method keys():

In [3]: iris.keys()

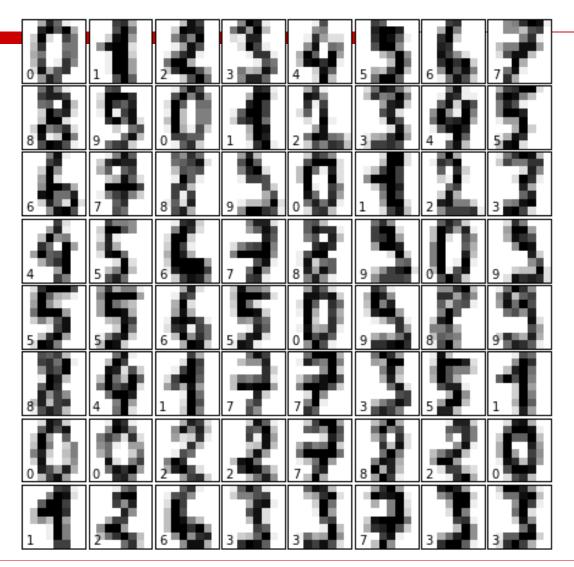
Out[3]: dict_keys(['target_names', 'data', 'feature_names', 'DESCR', 'target'])
```





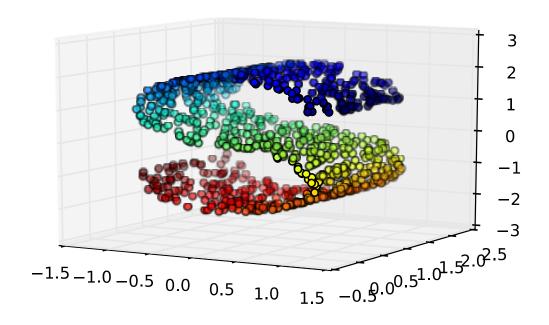


Digits

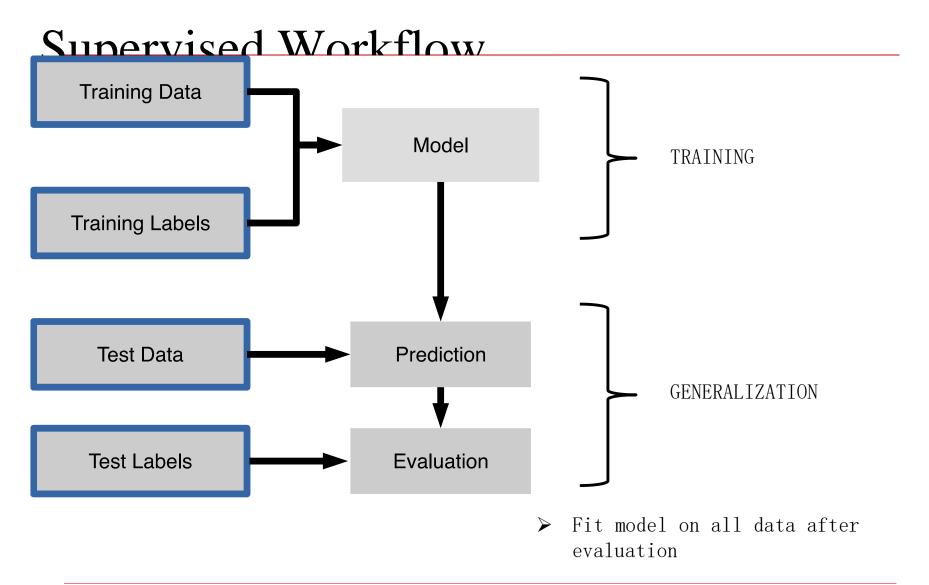




Generating Synthetic Data from sklearn. datasets import make_...







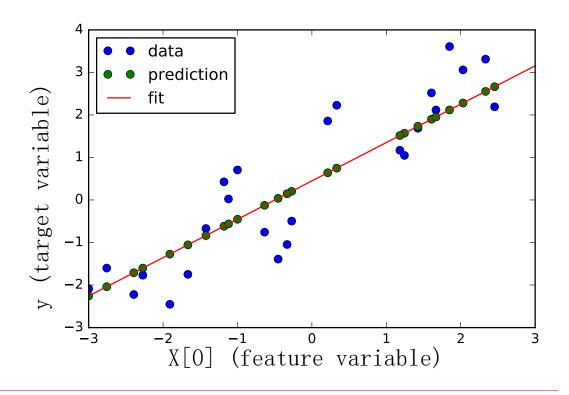


Supervised Workflow estimator. fit (X_train, y_train) **Training Data** TRAINING Model **Training Labels** estimator.predict(X_test) **Test Data** Prediction **GENERALIZATION Test Labels Evaluation**

estimator.score(X_test, y_test)

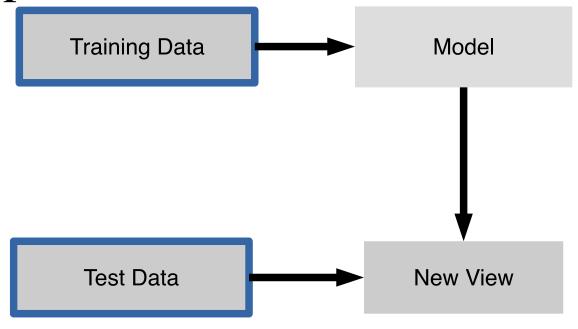


Linear Regression y =coef_[0]*X[0] + intercept_





Unsupervised Transformers



- ① transformer.fit(X_train)
- ② X_train_transf = transformer.transform(X_train)
- ③ X_test_transf = transformer.transform(X_test)



$$x_{std}^{(i)} = \frac{x^{(i)} - \mu_x}{\sigma_x}$$

 $x_{norm}^{(i)} = \frac{x^{(i)} - \mathbf{x}_{min}}{\mathbf{x}_{max} - \mathbf{x}_{min}}$

standardizatio

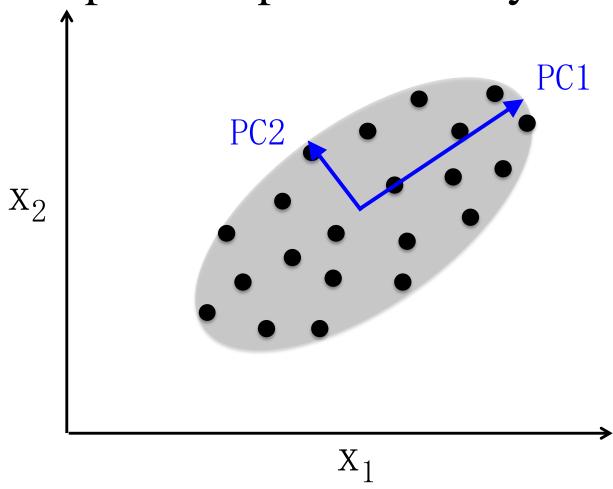
 η

	input	standardized	normalized
0	0	-1.46385	0.0
1	1	-0.87831	0.2
2	2	-0.29277	0.4
3	3	0.29277	0.6
4	4	0.87831	0.8
5	5	1.46385	1.0

min-max scaling
"normalization")

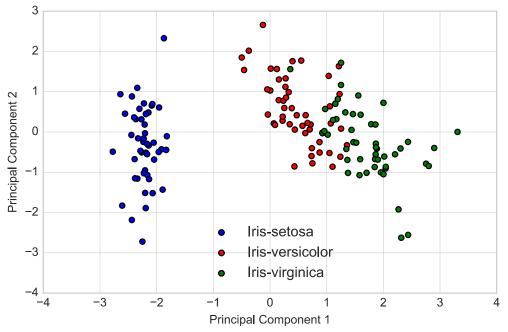


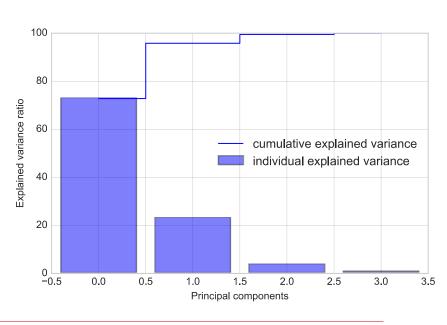
Principal Component Analysis





PCA for Dimensionality Reduction

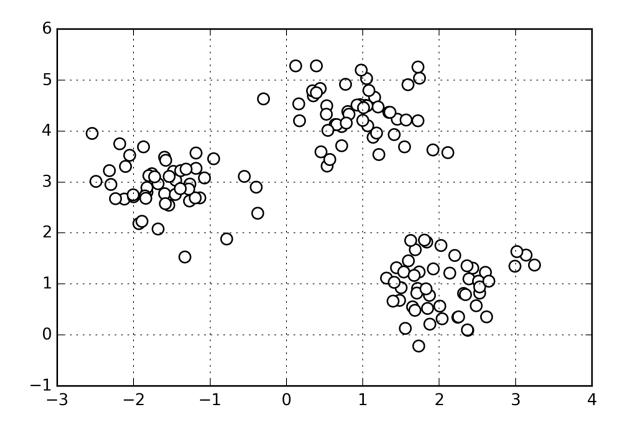






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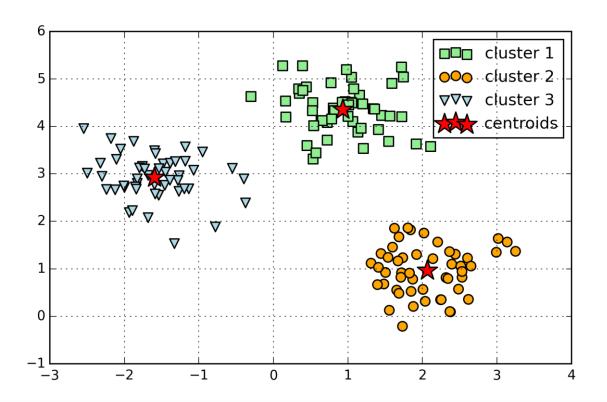
K-means Clustering





K-means Clustering

$$d(x,y)^{2} = \sum_{j=1}^{m} (x_{j} - y_{j})^{2} = ||x - y||_{2}^{2}$$





Scikit-learn API

<pre>estimator.fit(X_train, [y_train])</pre>							
<pre>estimator.predict(X_test)</pre>	<pre>estimator.transform(X_test)</pre>						
Classification	Preprocessing						
Regression	Dimensionality Reduction						
Clustering	Feature Extraction						
	Feature selection						



Bag of Words

• D1: "Each state has its own laws."

• D2: "Every country has its own culture."

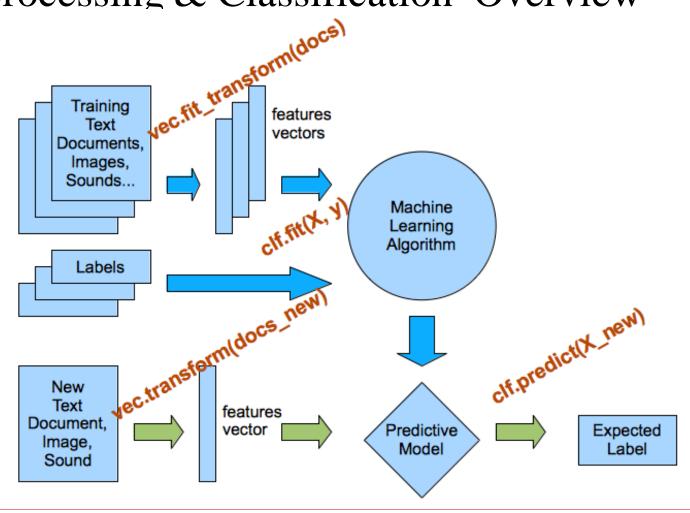
 $V = \{each: 1, state: 1, has: 2, its: 2, own: 2,$

laws: 1, every: 1, country: 1, culture: 1}

	each	state	has	its	own	laws	every	country	culture
\mathbf{x}_{D1}	1	1	1	1	1	1	0	0	0
\mathbf{x}_{D2}	0	0	1	1	1	0	1	1	1
\sum	1	1	2	2	2	1	1	1	1

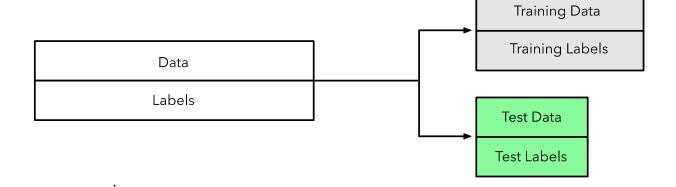


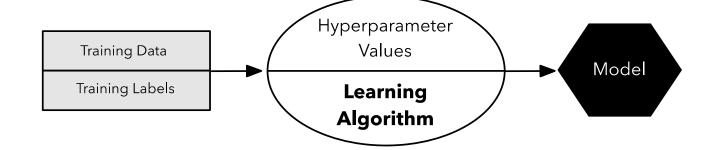
Preprocessing & Classification Overview



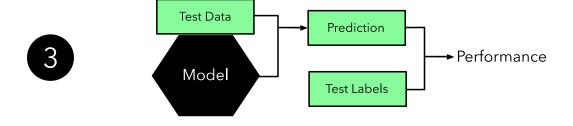


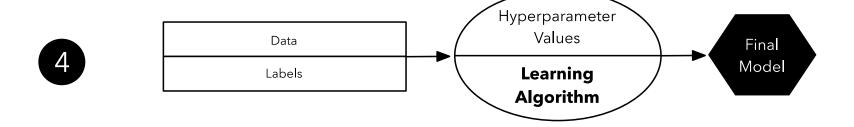
Holdout Evaluation I





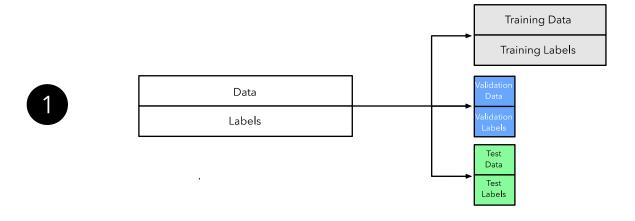
Holdout Evaluation II

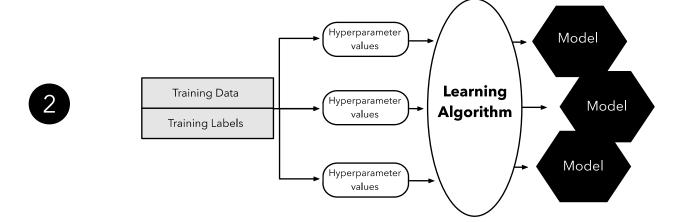




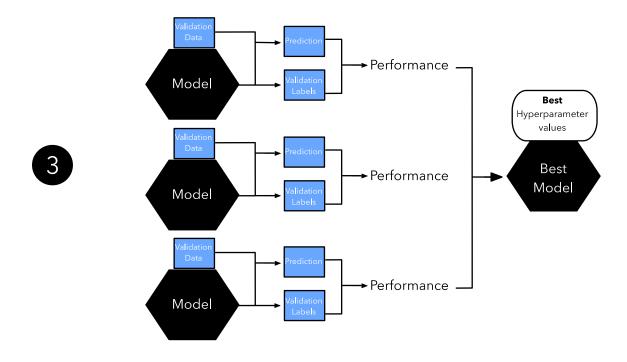
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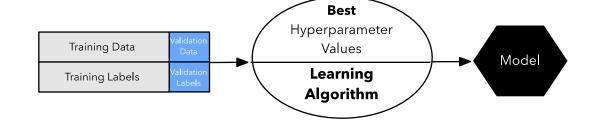
Holdout Validation I



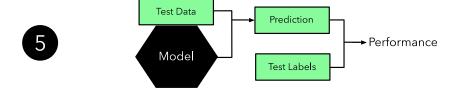


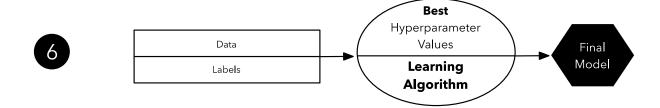
Holdout Validation II





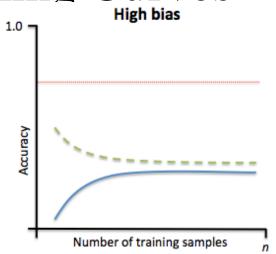
Holdout Validation III

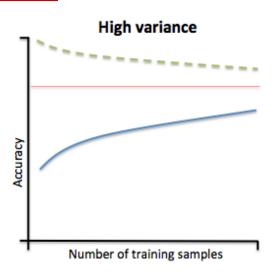


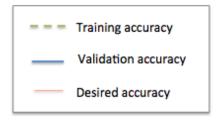


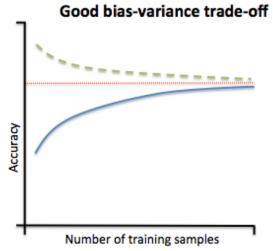


Learning Curves



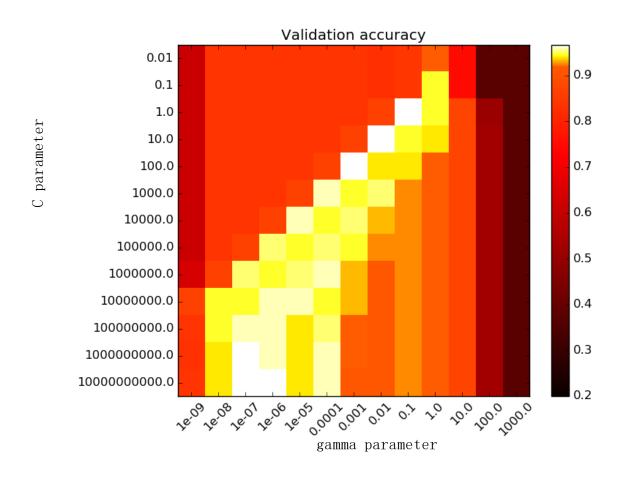








Grid Search





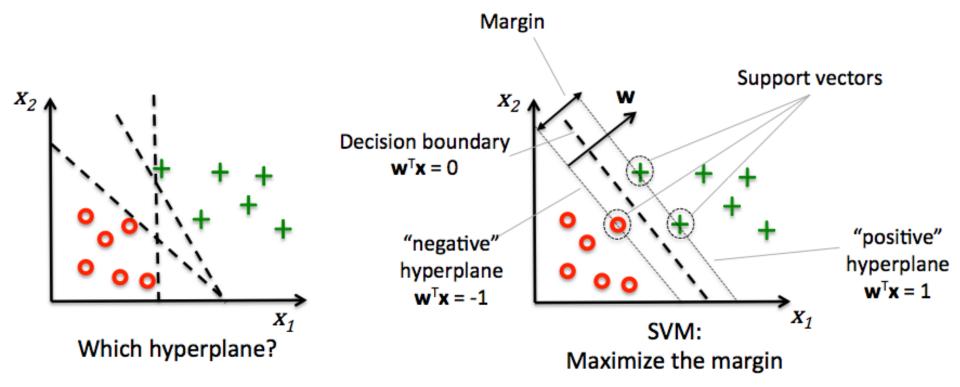
Confusion Matrix

Predicted class

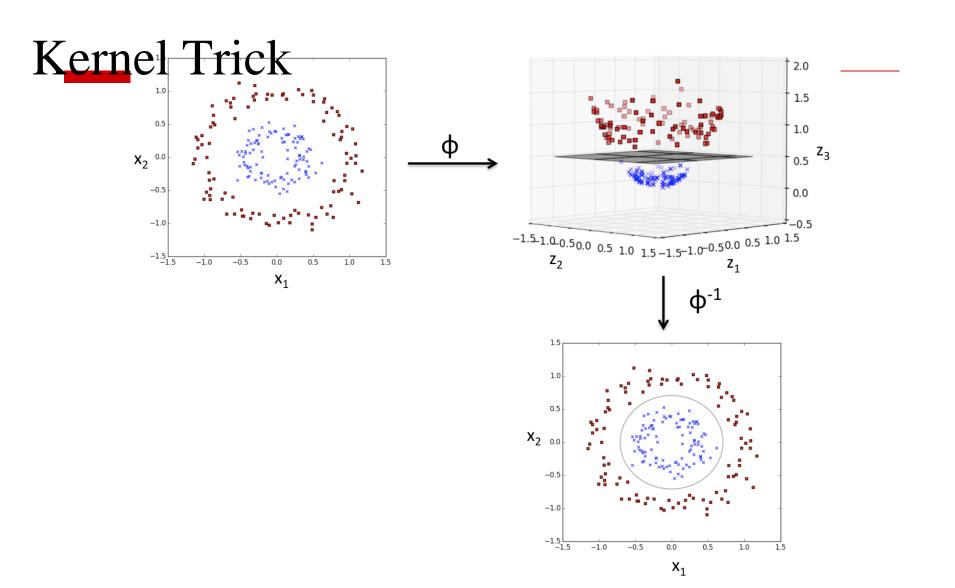
PNTrue False Positives Negatives (FN) **Actual** Class False True Positives NNegatives (TN)



Support Vector Machines

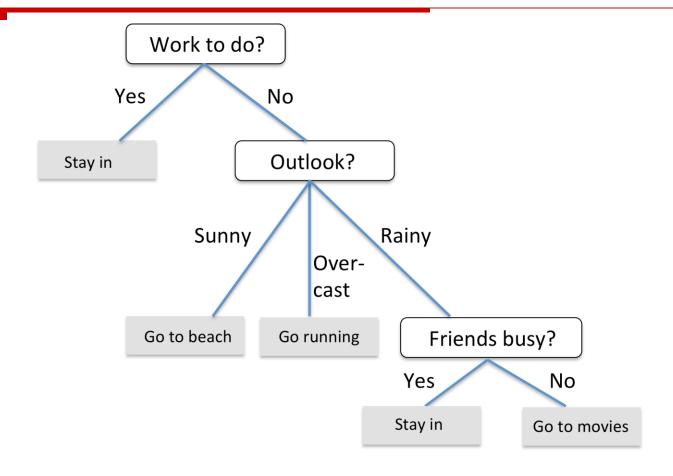






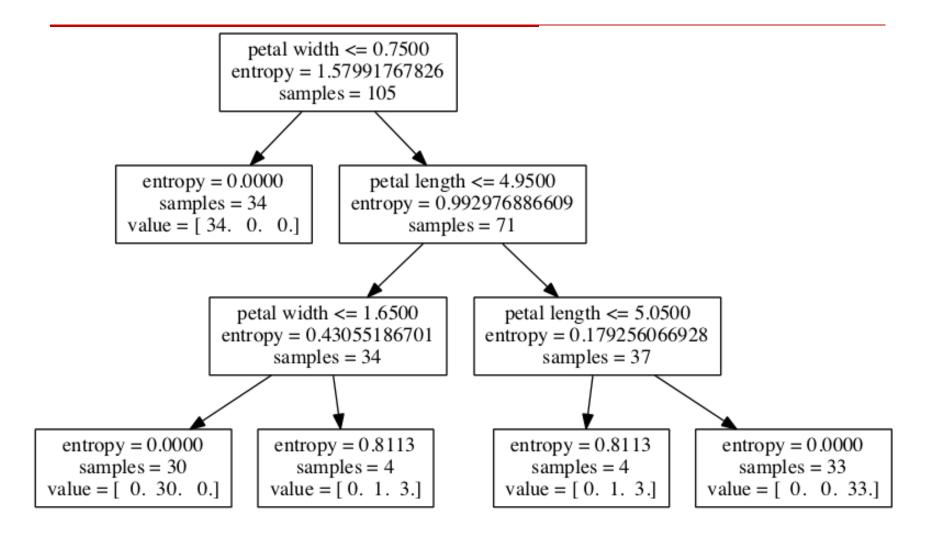


Decision Trees



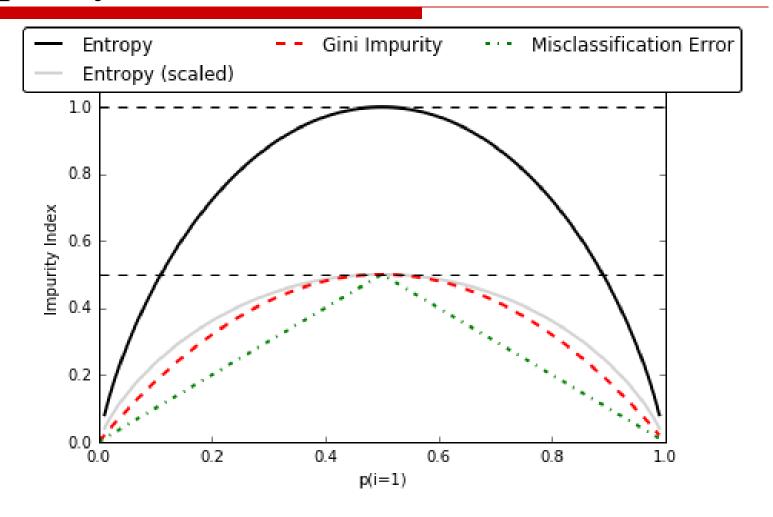


Classification w. Continuous Features





Impurity measures





Dimensionality Reduction

Dimensionality Reduction

Feature Selection

Feature Extraction



```
# Create first network with Keras
from keras.models import Sequential
from keras.layers import Dense
import numpy
# fix random seed for reproducibility
seed = 7
numpy.random.seed(seed)
# load pima indians dataset
dataset = numpy.loadtxt("pima-indians-diabetes.csv", delimiter=",")
# split into input (X) and output (Y) variables
X = dataset[:,0:8]
Y = dataset[:,8]
# create model
model = Sequential()
model.add(Dense(12, input dim=8, init='uniform', activation='relu'))
model.add(Dense(8, init='uniform', activation='relu'))
model.add(Dense(1, init='uniform', activation='sigmoid'))
# Compile model
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
# Fit the model
model.fit(X, Y, nb epoch=150, batch size=10)
# evaluate the model
scores = model.evaluate(X, Y)
print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
```



```
model = Sequential()
model.add(Convolution2D(nb_filters, kernel_size[0], kernel_size[1], border_mode='valid', input_shape=input_shape))
model.add(Activation('relu'))
model.add(Convolution2D(nb filters, kernel size[0], kernel size[1]))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=pool_size))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128))
model.add(Activation('relu'))
model.add(Dropout(0.5))
model.add(Dense(nb classes))
model.add(Activation('softmax'))
model.compile(loss='categorical_crossentropy',optimizer='adadelta',metrics=['accuracy'])
model.fit(X train, Y train, batch_size=batch_size, nb_epoch=nb_epoch,verbose=1, validation_data=(X test, Y test))
score = model.evaluate(X test, Y test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
```



```
import smtplib
from email.mime.text import MIMEText
mailto_list=["julyedu@qq.com,julyedu@126.com"]
mail host="smtp.126.com" #设置服务器
mail user="my126account" #用户名
mail pass="xiSm!x4" #口令
mail postfix="126.com" #发件箱的后缀
def send_mail(to_list,sub,content): #to list: 收件人; sub: 主题; content: 邮件内容
   me="yourname"+"<"+mail user+"@"+mail postfix+">"
   msg = MIMEText(content, subtype='html', charset='gb2312') #创建一个实例,这里设置为html格式邮件
   msg['Subject'] = sub #设置主题
   msg['From'] = me
   msg['To'] = ";".join(to_list)
   s = smtplib.SMTP()
   s.connect(mail host) #连接smtp服务器
   s.login(mail user, mail pass) #登陆服务器
   s.sendmail(me, to_list, msg.as_string()) #发送邮件
   s.close()
content = "test test 123"
send mail(to list, sub, content)
```



Selenium 模块实战

Demo时间。。

