CAPP 30122March 8

OUTSIDE TRADER

Hyun Ki Kim
Jessica Jee Yoon Song
Weiwei Zheng

Outside trader

- 1.Trading strategy: {"Ki": "1st year MACSS"}
 - Motivation
 - Web scraping
- 2. Modeling: {"Weiwei": "1st year MACSS"}
 - Machine learning
 - Model selection
- 3. Demo: {"Jessica": "1st year CAPP"}
 - Django

Motivation

Goal:

Building an algorithmic trading software

Insider trading:

- Easiest way to make money
- Illegal to reveal price sensitive information
- Secret may be revealed by collective search

Search ranking:

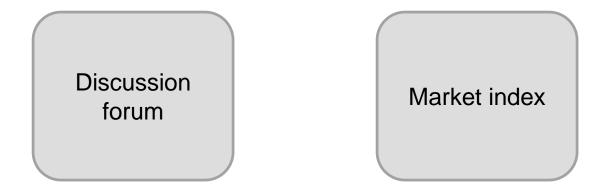
- Preceding indicator for stock price
- Need real time web scraping
- Naver Finance (Korea-based web portal)

Web scraping

Filtering process (overnight)



Real-time scraping (9:01am ~ 3:30 pm)



CRON

APScheduler

Executes job function periodically

```
from apscheduler.schedulers.blocking import BlockingScheduler
def job_function:
    define the task you wanna run.
    print ("hello world!")
    return
sched = BlockingScheduler()
# set the parameters as you want
sched.add_job(job_function, 'cron', month='2', day='7-28', hour='9-12', \
              minute='0-59/15')
sched.start()
```

Difficulties

```
Execution of job "job function (trigger: cron[month='3', day='1-/, noul- 0-2
-03-01 18:40:00 CST) " skipped: maximum number of running instances reached (1
Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-2:
-03-01 18:50:00 CST) " skipped: maximum number of running instances reached (1)
Execution of job "job_function (trigger: cron[month='3', day='1-7', hour='0-23
 -03-01 19:00:00 CST) " skipped: maximum number of running instances reached (1)
 Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23
 -03-01 19:10:00 CST) " skipped: maximum number of running instances reached (1)
 Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23
 -03-01 19:20:00 CST) " skipped: maximum number of running instances reached (1)
 Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23
 -03-01 19:30:00 CST) skipped: maximum number of running instances reached (1)
 Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23
  -03-01 19:40:00 CST) " skipped: maximum number of running instances reached (1)
  Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23
  -03-01 19:50:00 CST) " skipped: maximum number of running instances reached (1)
  Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23'
  -03-01 20:00:00 CST) " skipped: maximum number of running instances reached (1)
  Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23'
  -03-01 20:10:00 CST)" skipped: maximum number of running instances reached (1)
   Execution of job "job function (trigger: cron[month='3', day='1-7', hour='0-23'
```

Difficulties

KOSDAQ Ind Period 10	CX	Felate +3.17 768,912,000 95) Compare 18:30 - 01:35 Lst	0 846.	40 H 861.08 ons • 97) Ed
1D 3D 1M	6M YTD 1Y	5Y Max 10 Min	THE RESERVE AND ADDRESS OF THE PERSON NAMED IN	Mov Avgs
Date		KOSDAQ Ind		▼ Chart
02/18 19:00	Open Open	High	Low	Close
02/14 01:30	855.78	856.88	855.58	856.81
02/14 01:30	846.89	848.03	846.89	848.03
The state of the s	846.56	846.56	846.56	846.56
02/14 01:10	845.05	846.56	844.98	846.56
02/14 01:00	843.20	845.45	843.20	844.88
02/14 00:50	843.48	843.48	841.85	843.14
02/14 00:40	842.08	843.53	841.38	843.53
02/14 00:30	843.69	843.91	842.32	842.32
02/14 00:20	841.58	843.70	841.58	843.61
02/14 00:10	841.48	841.73	840.81	841.57
02/14 00:00	841.45	841.51	839.69	841.51
02/13 23:50	840.28	841.59	840.25	841.38
02/13 23:40	838.05	840.48	838.05	840.32

Modeling

Data

variables

Method

- machine learning
- model selection

Results

- performance of best models
- next step

Variables

Dependent variable

- predicted if price increases by more than 0.33% after an hour
- made prediction every ten minutes

Predictors (in total 164)

use data at the moment, ten and twenty minutes earlier

Discussion

trend of posts number, unique participants and click, etc.

Price

closing price the day before, price volatility, price trend, difference between selling and buying price of each period, price gap volatility, trade volume trend, etc.

Variables

Dependent variable

- predicted if price increases by thirty percent one hour later
- made prediction every ten minutes

Predictors (in total 164)

use data at the moment, ten and twenty minutes earlier

<u>Time</u>

different time slots (early morning, morning, lunch, etc.)

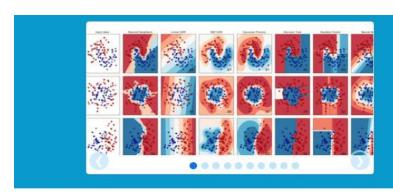
Market index

KOSPI and KOSDAQ index and their changes

Different square and interaction terms

Predicting methods

scikit-learn API



scikit-learn

Machine Learning in Python

- Simple and efficient tools for data mining and data analysis
- Accessible to everybody, and reusable in various contexts
- · Built on NumPy, SciPy, and matplotlib
- · Open source, commercially usable BSD license

Classification

Identifying to which category an object belongs to.

Applications: Spam detection, Image recognition.

Algorithms: SVM, nearest neighbors, random forest. ... — Examples

Regression

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices.

Algorithms: SVR, ridge regression, Lasso, ...

- Examples

Clustering

Automatic grouping of similar objects into sets.

Applications: Customer segmentation, Grouping experiment outcomes

Algorithms: k-Means, spectral clustering, mean-shift. ... — Examples

Dimensionality reduction

Reducing the number of random variables to consider.

Applications: Visualization, Increased efficiency

Algorithms: PCA, feature selection, nonnegative matrix factorization. — Examples

Model selection

Comparing, validating and choosing parameters and models.

Goal: Improved accuracy via parameter tuning

Modules: grid search, cross validation, metrics. — Examples

Preprocessing

Feature extraction and normalization.

Application: Transforming input data such as text for use with machine learning algorithms. **Modules**: preprocessing, feature extraction.

Activate Windows

Examples

Go to Settings to activate Windows

Predicting methods

Logistic regression

from sklearn.linear_model import LogisticRegression

K Nearest Neighbors (KNN)

from sklearn.neighbors import KNeighborsClassifier

Principal component regression (PCR)

from sklearn.decomposition import PCA

Partial least squares regression (PLS)

from sklearn.cross_decomposition import PLSRegression

Tree-based methods

Suport vector machines (SVM)

from sklearn.svm import SVC

Model selection

Validation set approach

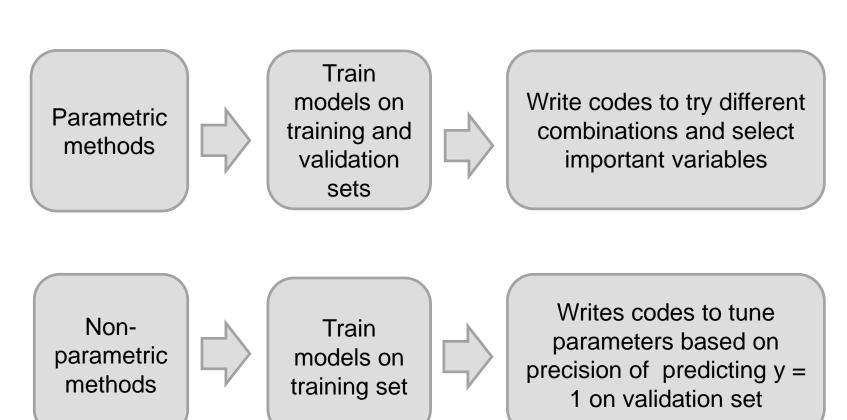
- training set: February 14, 20, 21
- validation set: February 22, 23, 26
- testing set: February 27, 28, March 2, 3, 6, 7
- Observations in total: 19786

Tuning process

- Regression vs non-linear models
- Penalty for non-linearity
- Using mean square error
 from sklearn.metrics import mean_squared_error, classification_report

Model selection

Fitting process



Tree-based methods

decision tree, bagging, random forest, boosting

Decision Tree

- from sklearn.tree import DecisionTreeClassifier, export_graphviz
- import graphviz

Important variables

price volatility: (maximum price – minimum price) / minimum price

average price volatility: price volatility / increase count

alpha: price increase – market increase

<u>variation ten minutes before squared</u>: the squared price variation between ten minutes before and eleven minutes before

Tree-based methods

