KDD Cup 2017

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Problem Description

TASK 1:

Travel Time Prediction

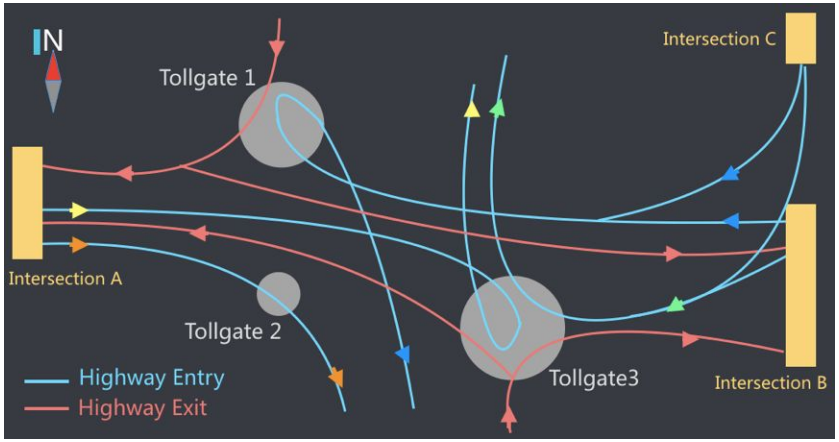
For every 20-minute time window, estimate the average travel time of each route.

A. Intersection A - Tollgates 2 & 3
B. Intersection B - Tollgates 1 & 3
C. Intersection C - Tollgates 1 & 3

TASK 2:

Traffic Volume Prediction

For every 20-minute time window, predict the entry and exit traffic volumes at tollgates 1, 2 and 3 from Oct. 25th - Oct. 31st.



Competition website : [KDD Cup 2017]

How To Do?

Day	startTime	x1	x2	h	y
1	1	71	103	127	118
1	7	90	115	90	86
1	2	103	118	127	168
1	8	115	86	90	85
1	3	118	168	127	161
1	9	86	85	90	91
1	4	168	161	127	145

TASK 2

```
1  Label date with 1-7 which means Mon. to Sun.
2  Label time with 1-12(1: 8:00-8:20; 7: 17:00-17:20)
3  Use sqlQuery to count volume of each time range(every 20 mins)
4  Get average volume of 8-10 and 17-19 of each day
5  Create training data with day label, time, volumes of two previous time
•  range, average volume of the same time range it belongs to(8-10 or 17-19)
6  Regress with Linear Regression
7  Fit the model
8  Predict volume
```

Previous Result

Linear Regression **0.2539**

Feature:

- What day?
- Time range
  - One unit / 20 min
  - 8-10, 17-19
- Volume of two previous time range
  - 8:00-8:20 => 7:20-7:40, 7:40-8:00
- Average volume of the same time range(8-10 or 17-19) of that day

Grade

Volume Prediction 592 / 0.2220

Travel Time Prediction | Volume Prediction

时间	MAPE	当天排名
2017-05-24 12:49:30	0.2220 ↑	148

Volume Prediction 427 / 0.4213

Travel Time Prediction | Volume Prediction

时间	MAPE	当天排名
2017-06-01 13:44:56	0.4213 ↑	229
2017-05-28 13:28:27	0.4248 ↓	274

Using Xgboost

n_estimators=10

- 0.2457

n_estimators=20

- 0.2401

n_estimators=100

- 0.2451

Using xgboost method surely improve the result. First, we try different n_estimators.

Using Xgboost

linear booster; n_estimator=50

- 0.2500

tweedie

- 0.2550

gamma

- 0.2665

According to the result, n_estimators=20 when using linear booster got the best result.

Using Xgboost

learning_rate=0.55

- 0.2248

learning_rate=0.6

- 0.2220

learning_rate=0.65

- 0.2295

According to the result, when reducing learning_rate, the result might be better.

Briefly conclusion for Task 2

- We can infer that the model fall into overfitting/underfitting problem from our experiments. (Overfitting, estimator >= 50; Underfitting estimator = 10.) Therefore, we chose 20.

- Second, we chose differents algorithms, including Linear Regression, Gamma, Tweedie. Linear Regression is better than others, according to results.

- Next, we modified learning rate for our model, get 0.6 is better than 5.5 and 6.5.

- Finally, we get 0.2220 grade by using Linear Regression with estimators 20 and 0.6 learning rate in our xgboost parameters.

TASK 1

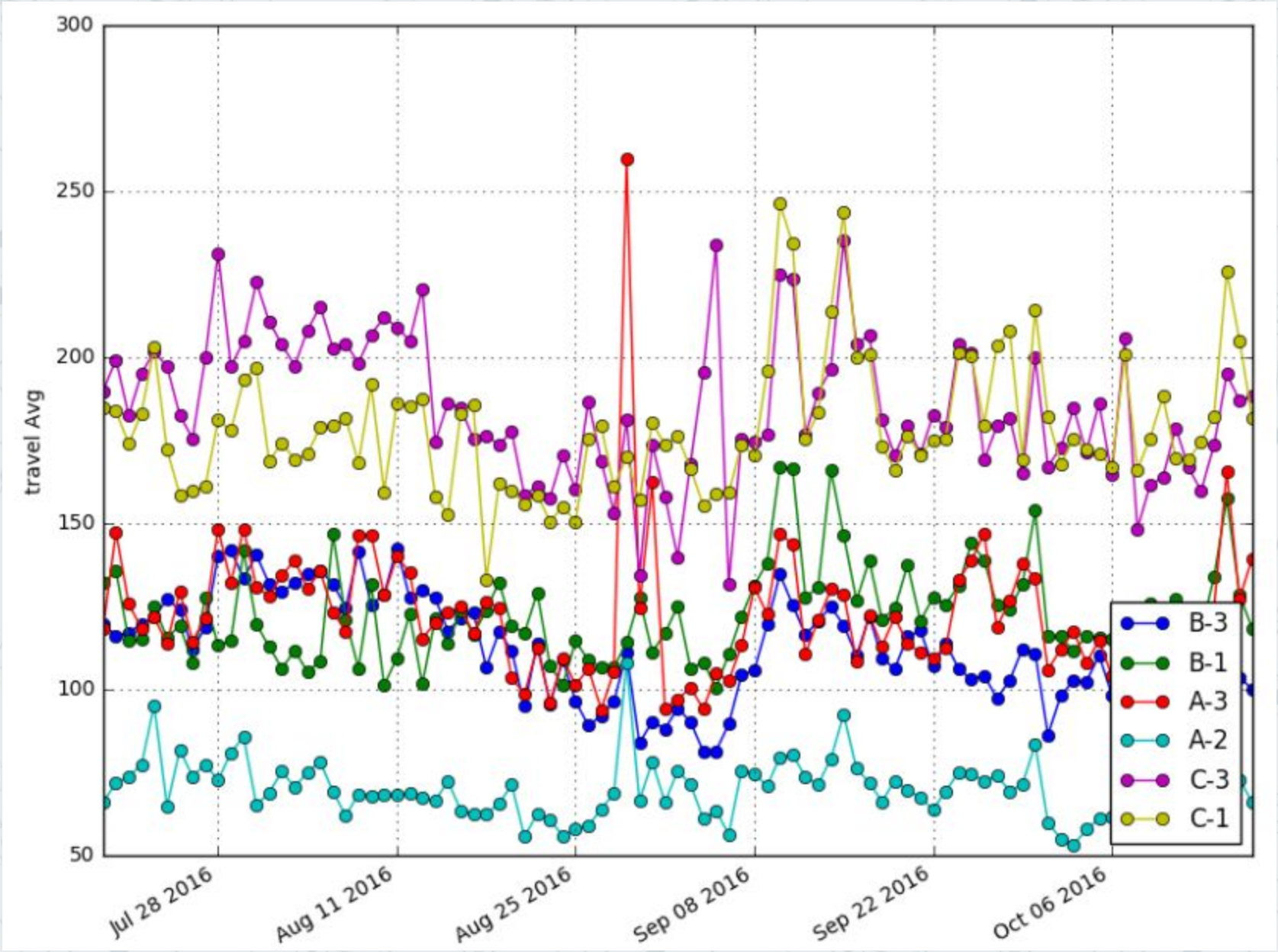
```
Get 21 Models by dividing intersection_id and day(Mon.-Sun.)
Address noise data by ignoring the data whose travel_time is less than
  25% and higher than 75%
For i in 21 Models:
  Label starting_time
  Train with DecisionTreeRegressor with AdaBoostRegressor
  Predict travel_time
```

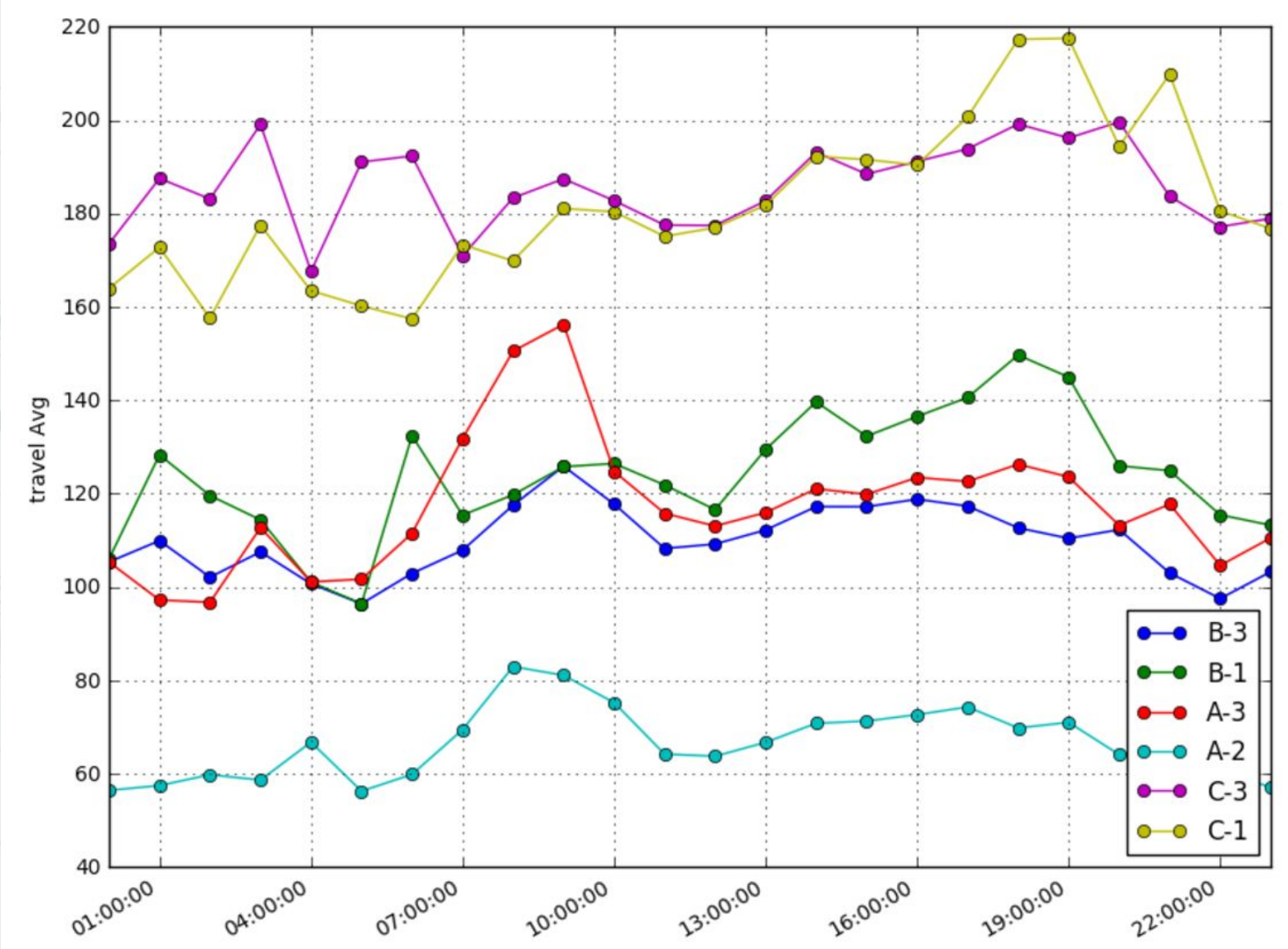
Grade

Travel Time Prediction 217 / 0.2045

Travel Time Prediction | Volume Prediction

时间	MAPE	当天排名
2017-06-01 13:54:40	0.2045 ↓	111
2017-05-31 14:37:36	0.2045 ↓	127


300
250
200
150
100
50
travel Avg
Jul 28 2016
Aug 11 2016
Aug 25 2016
Sep 08 2016
Sep 22 2016
Oct 06 2016
B-3
B-1
A-3
A-2
C-3
C-1


travel Avg
220
200
180
160
140
120
100
80
60
40
01:00:00
04:00:00
07:00:00
10:00:00
13:00:00
16:00:00
19:00:00
22:00:00
B-3
B-1
A-3
A-2
C-3
C-1

Data Preprocessing

- Divide data into 21 sets
  - with intersection id (A, B, C)
  - what day? (Mon. - Sun.)
- Label Mon.- Sun. with 0-7

Training models => 21 models (3 * 7)

intersection A - Mon.

intersection A - Tue.

intersection A - Wed.

…

intersection C - Sun.

intersection_id	tollgate_id	vehicle_id	starting_time	travel_seq	travel_tim	weekend
A	2	1014410	2016/7/25 00:16	110#2016-07-25 00	41.39	0
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						

intersection_id	tollgate_id	vehicle_id	starting_time	travel_seq	travel_tim	weekend
A	2	1071181	2016/7/19 00:37	110#2016-07	58.05	1
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						
A						

intersection_id	tollgate_id	vehicle_id	starting_time	travel_seq	travel_tim	weekend
C	3	1064323	2016/7/24 01:29	115#2016-07-	222.53	6
C	1	1018237	2016/7/24 05:36	115#2016-07-	186.77	6
C	1	1056942	2016/7/24 06:00	115#2016-07-	158.49	6
C	1	1079264	2016/7/24 06:29	115#2016-07-	126.98	6
C	1	1034865	2016/7/24 06:52	115#2016-07-	154.93	6
C	1	1018948	2016/7/24 07:01	115#2016-07-	123.44	6
C	3	1024998	2016/7/24 07:04	115#2016-07-	39.35	6
C	3	1002743	2016/7/24 07:20	115#2016-07-	119.64	6
C	3	1072724	2016/7/24 07:20	115#2016-07-	188.21	6
C	1	1054062	2016/7/24 07:22	115#2016-07-	219.5	6
C	3	1071321	2016/7/24 07:25	115#2016-07-	210.02	6
C	1	1007882	2016/7/24 08:00	115#2016-07-	157.03	6
C	3	1063882	2016/7/24 08:17	115#2016-07-	207.95	6
C	1	1064065	2016/7/24 08:19	115#2016-07-	146.88	6

Problem - Travel Time

A-2

mean	70.123898
std	45.561928
min	9.260000
25%	44.980000
50%	58.660000
75%	82.715000
max	1569.640000

A-3

mean	123.824527
std	83.335008
min	19.790000
25%	88.860000
50%	107.710000
75%	137.210000
max	6711.110000

B-1

mean	128.078528
std	57.578811
min	19.460000
25%	96.290000
50%	117.850000
75%	144.510000
max	1627.380000

B-3

mean	113.412535
std	53.858812
min	11.740000
25%	78.700000
50%	106.315000
75%	137.942500
max	1498.970000

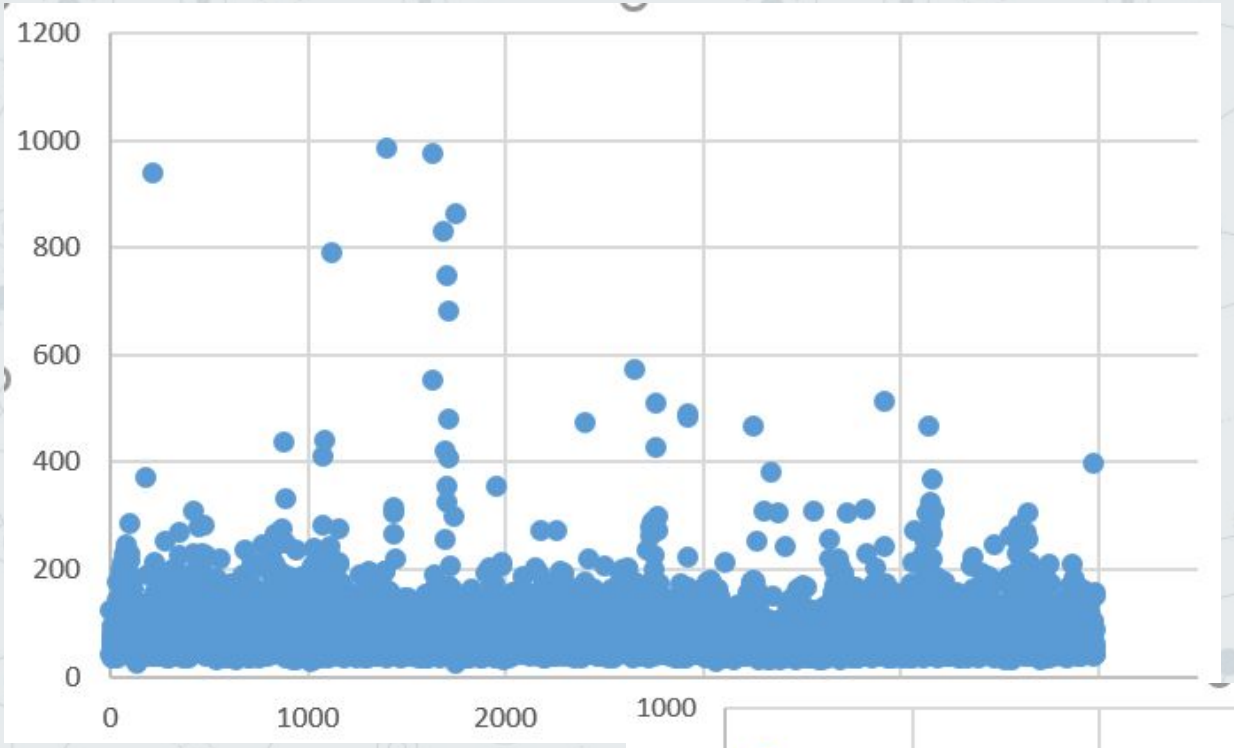
C-1

mean	184.307117
std	73.699985
min	38.500000
25%	142.140000
50%	171.455000
75%	210.382500
max	2489.570000

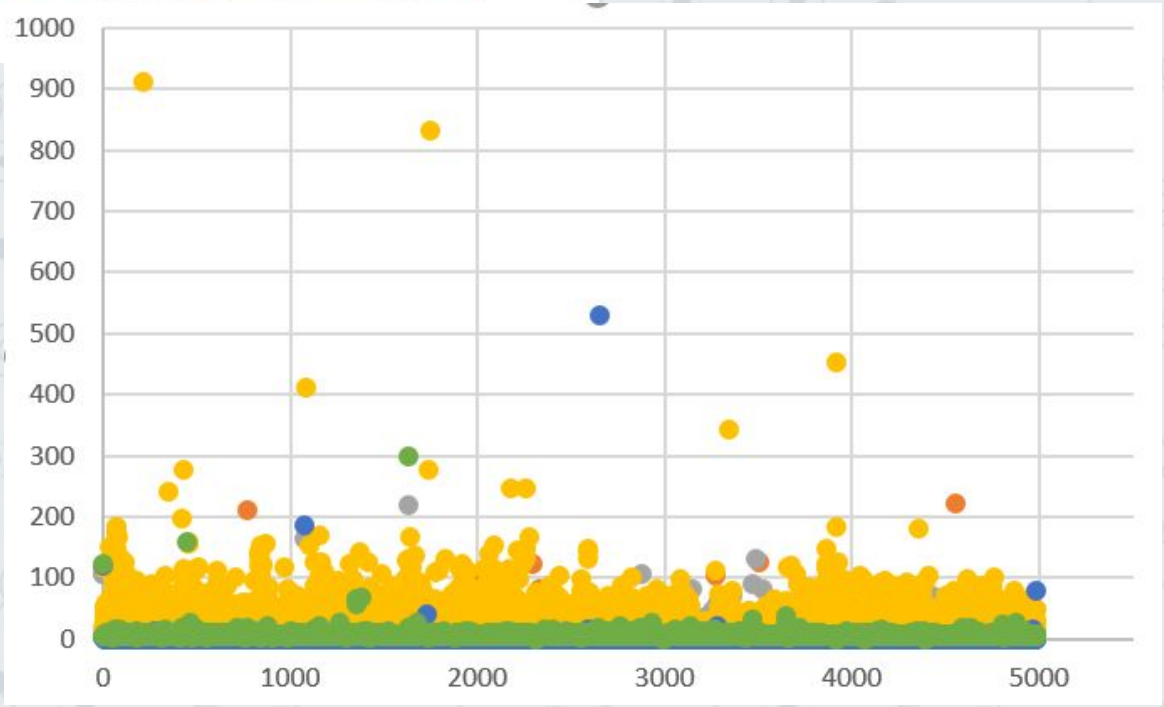
C-3

mean	187.242564
std	72.014020
min	32.040000
25%	142.830000
50%	176.200000
75%	217.170000
max	1260.760000

107	108	110	117	120	123	intersection_id	starting_time	tollgate_id	travel_time
3.04	3.57	9.42	19.39	0.54	5.27	A	2016/7/25 00:16	2	41.39
3.26	3.83	24.77	0	0	5.65	A	2016/7/25 00:46	3	122.91
3.45	4.05	25.2	27.79	0.61	5.98	A	2016/7/25 01:55	2	66.79
3.9	4.59	16.52	17.03	0.69	7.62	A	2016/7/25 02:08	2	51.03
4.19	8.52	22.75	52.59	1.36	4.34	A	2016/7/25 02:28	2	93.59
3.33	3.92	8.4	13.33	0.59	5.78	A	2016/7/25 03:24	2	35.33
4.23	4.97	13.55	16.91	0.75	7.34	A	2016/7/25 04:23	2	47.91
4.15	4.88	13.43	45.61	0.73	7.2	A	2016/7/25 04:36	2	75.61
4.38	5.15	9.12	36.75	0.77	7.6	A	2016/7/25 04:44	2	63.75
4	4.71	8.54	38.09	0.71	4.62	A	2016/7/25 04:45	2	61.09
3	3.53	9.62	59.34	0.53	5.21	A	2016/7/25 04:54	2	81.34
4.73	5.56	15.16	18.91	0.83	8.2	A	2016/7/25 04:55	2	52.91
3.53	4.16	15.29	20.7	0.62	6.13	A	2016/7/25 04:55	2	50.7



total travel time



travel time of each link

Remove Noise

- As the table shown, there exists noise in the data (e.g. maximum time, minimum time)
- As the figure shown, the distribution position of total time is similar to that of travel time of each link
- Take the data which total travel time is in the range of 25%-75% as training data

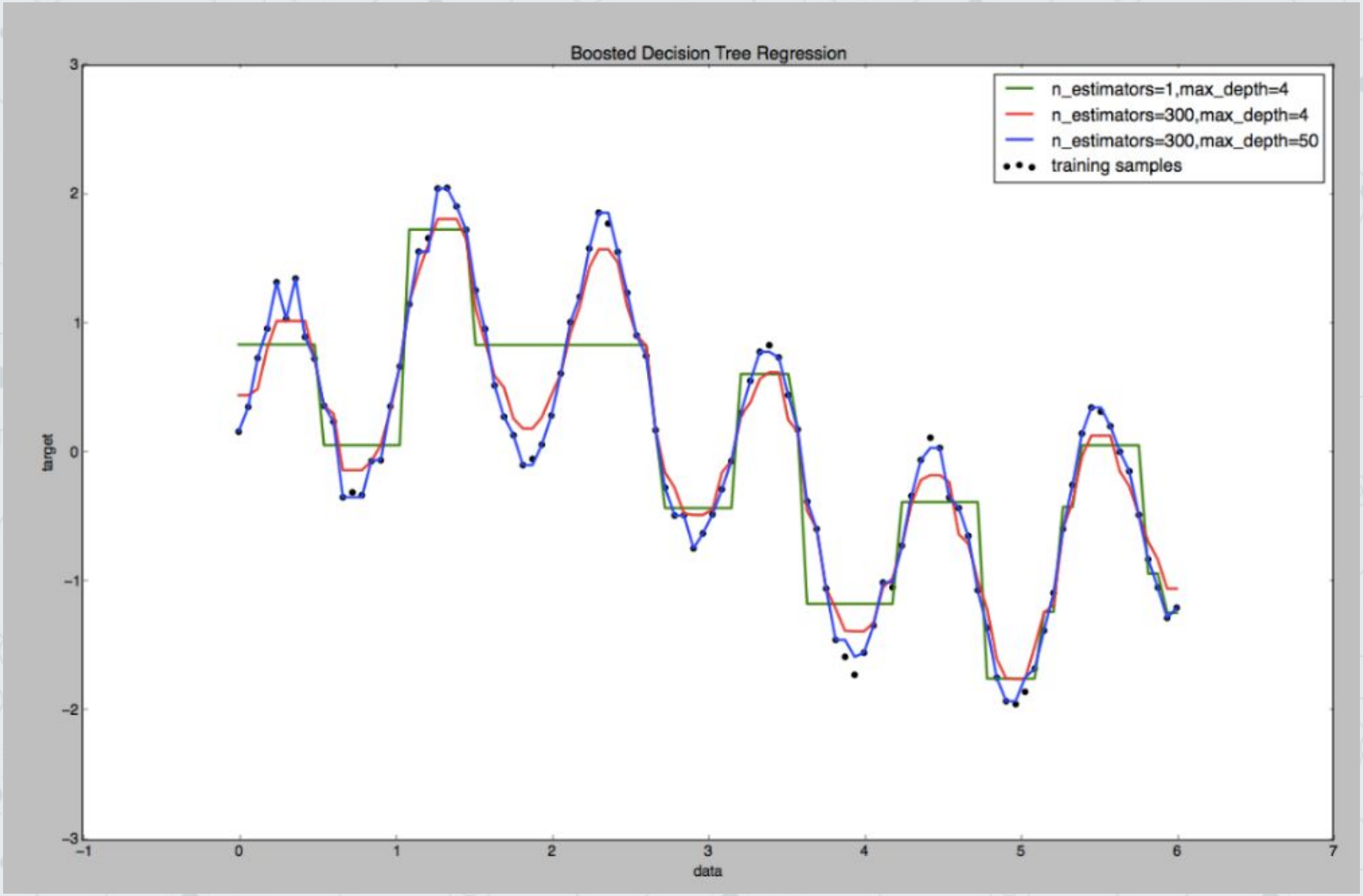
Decision Tree with Adaboost

Decision Tree

- Easily overfit

Adaboost

- Improve accuracy
- Avoid overfitting


Boosted Decision Tree Regression
n_estimators=1,max_depth=4
n_estimators=300,max_depth=4
n_estimators=300,max_depth=50
training samples
target
data
3
2
1
0
−1
−2
−3
−1
0
1
2
3
4
5
6
7

Parameters Selection

n_estimators

- 50
- 100
- 300

loss function

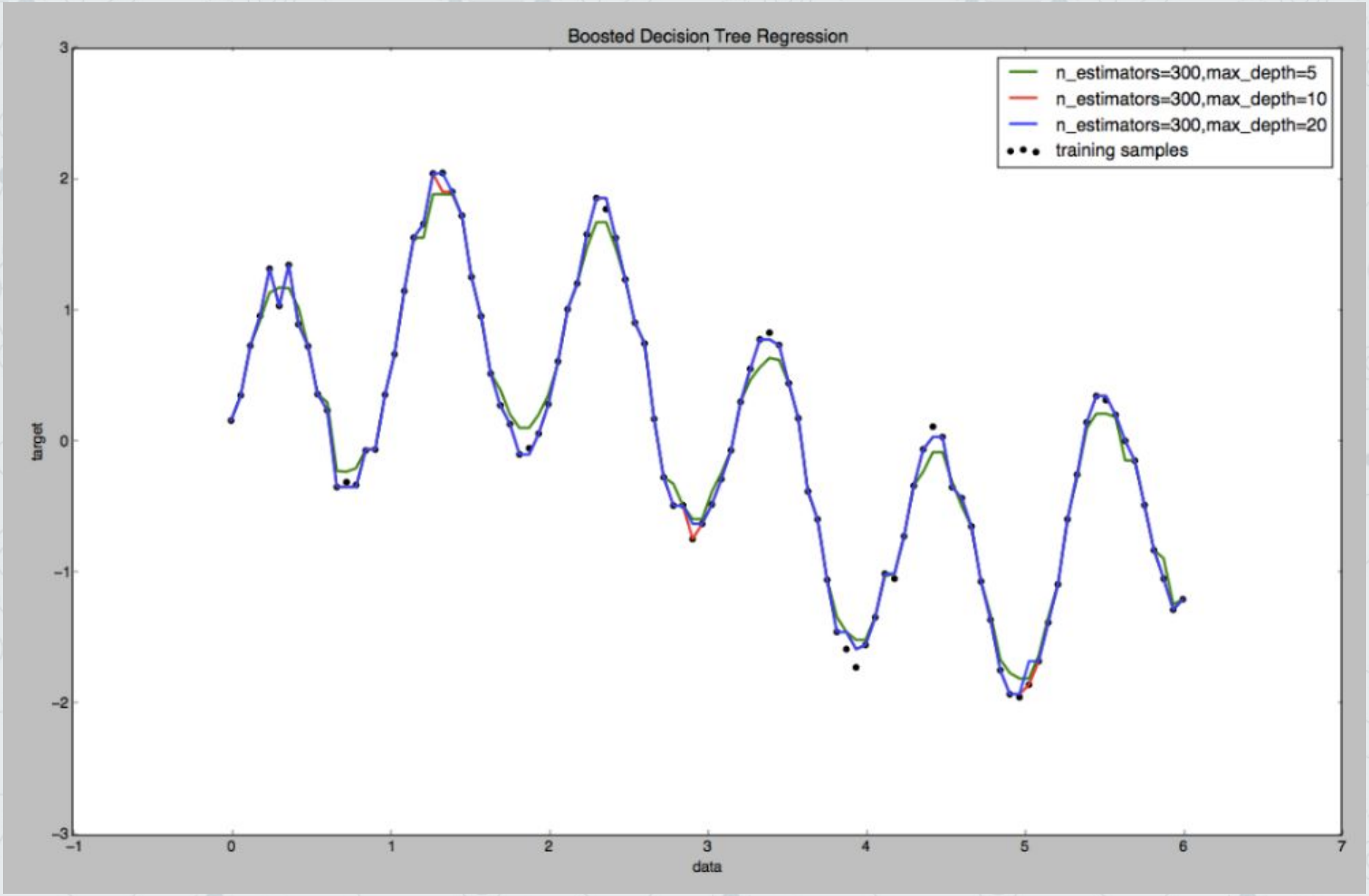
- linear
- square
- exponential

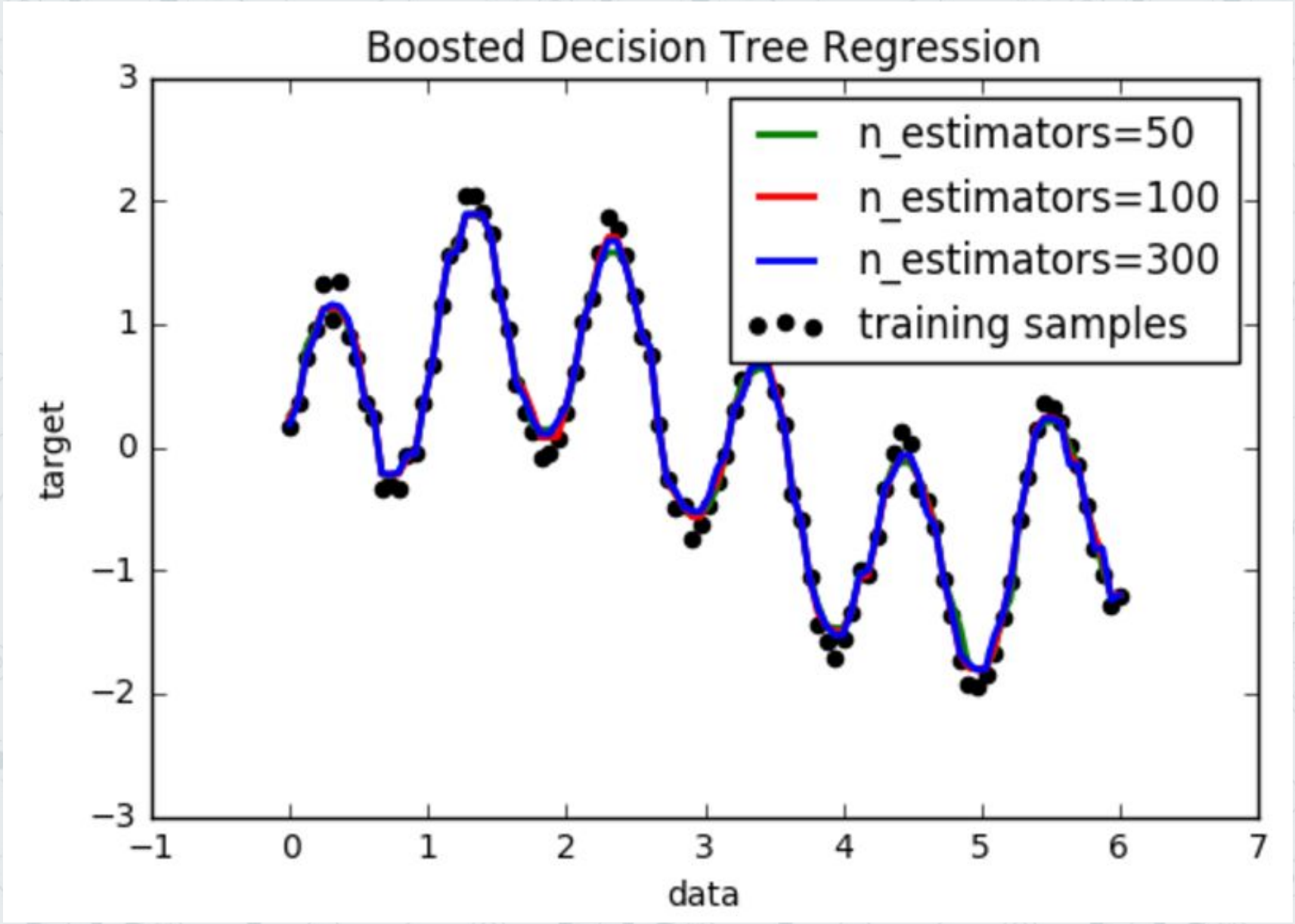
max_depth

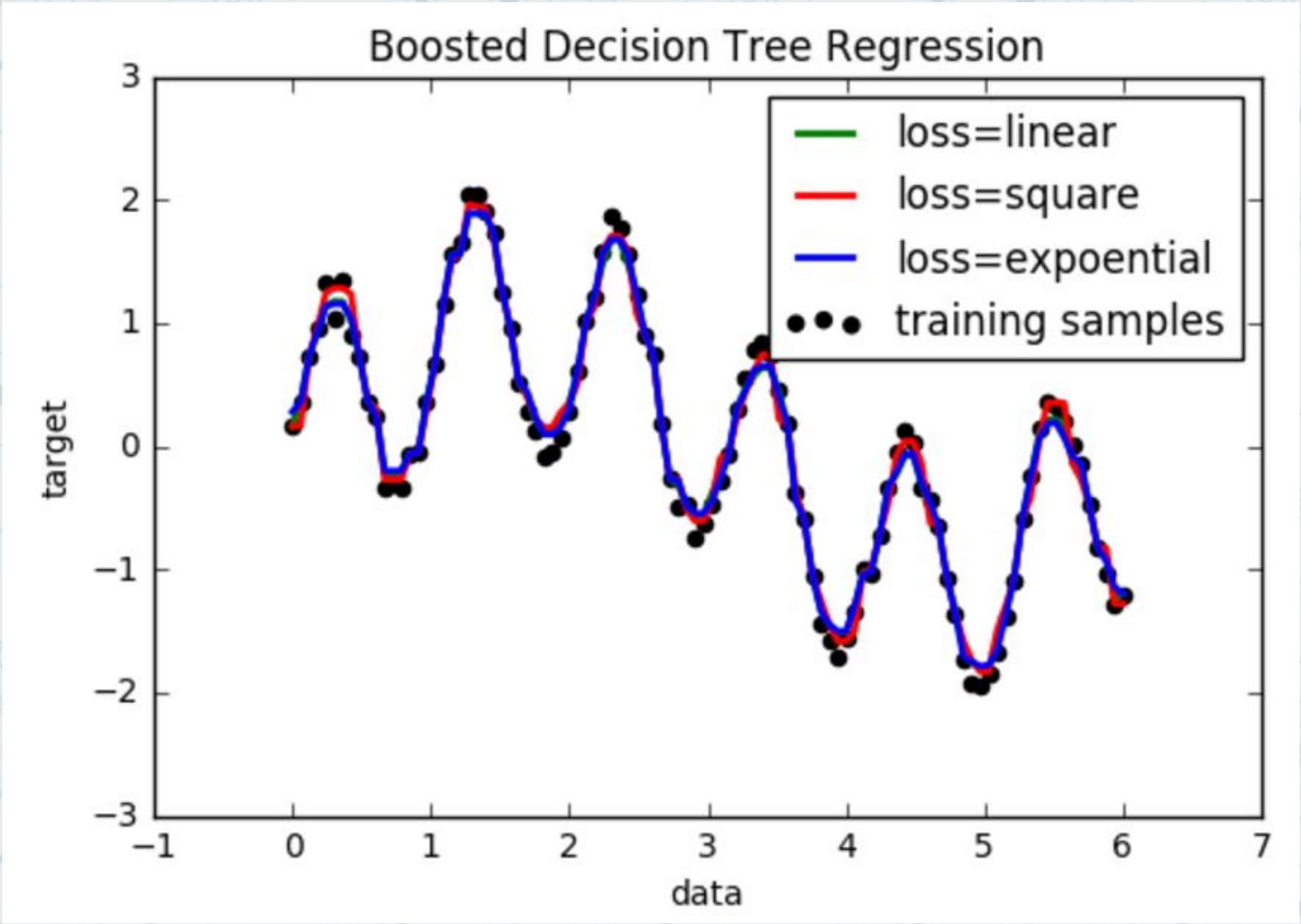
- 5
- 10
- 20

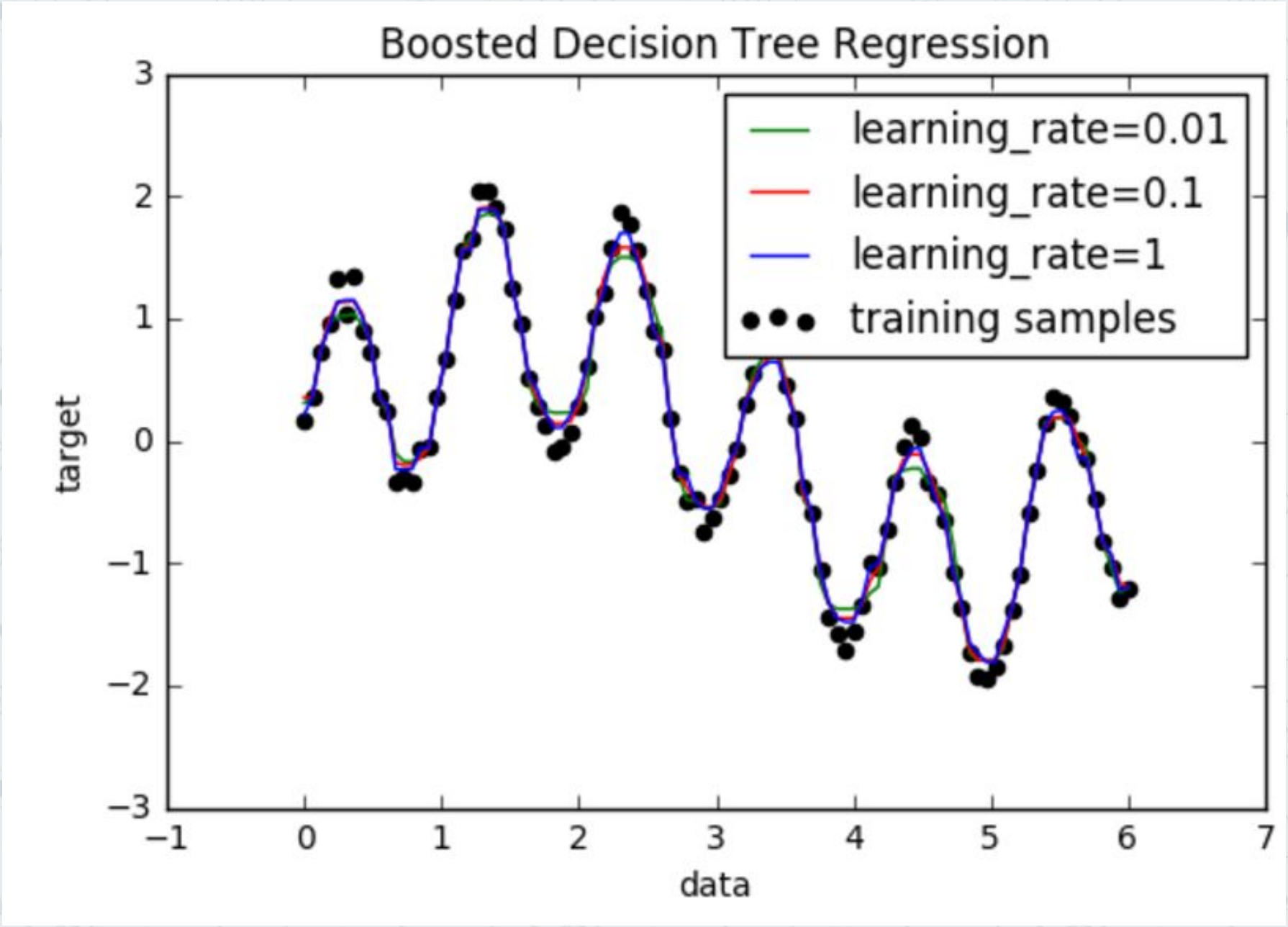
learning_rate

- 0.01
- 0.1
- 1


Boosted Decision Tree Regression
n_estimators=300,max_depth=5
n_estimators=300,max_depth=10
n_estimators=300,max_depth=20
training samples
target
data
3
2
1
0
−1
−2
−3
−1
0
1
2
3
4
5
6
7


Boosted Decision Tree Regression
n_estimators=50
n_estimators=100
n_estimators=300
training samples
target
data
3
2
1
0
−1
−2
−3
−1
0
1
2
3
4
5
6
7


Boosted Decision Tree Regression
loss=linear
loss=square
loss=expoential
training samples
target
data
3
2
1
0
−1
−2
−3
−1
0
1
2
3
4
5
6
7


Boosted Decision Tree Regression
learning_rate=0.01
learning_rate=0.1
learning_rate=1
training samples
target
data
3
2
1
0
−1
−2
−3
−1
0
1
2
3
4
5
6
7

Decision Tree Regression with Adaboost

Feature

- Start point time
- Exit path

Target

- Total travel time

Field	*Type*	*Description*
intersection_id	string	intersection ID
tollgate_id	string	tollgate ID
vehicle_id	string	vehicle ID
starting_time	datetime	time point when the vehicle enters the route
travel_seq	string	trajectory in the form of a sequence of link traces separated by ";", each trace consists of link id, enter time, and travel time in seconds, separated by "#"
travel_time	float	the total time (in seconds) that the vehicle takes to travel from the intersection to the tollgate

Table 5

starting_time	travel_seq	travel_time
2016/7/19 00:14	105#2016-	70.85
2016/7/19 00:35	105#2016-	148.79
2016/7/19 00:37	105#2016-	79.76
2016/7/19 00:37	110#2016-	58.05
2016/7/19 00:56	105#2016-	137.98
2016/7/19 00:56	115#2016-	113.54
2016/7/19 01:26	105#2016-	176.7
2016/7/19 01:36	110#2016-	74.47
2016/7/19 01:36	110#2016-	94.57
2016/7/19 01:36	115#2016-	214.87

Result

max_depth=5

- 0.2045

max_depth=10

- 0.2114

max_depth=20

- 0.2096

Although the more depth the tree is, the better the result we got when learning, it might be overfitting.

Get Model 2

Use previous model to train the second model (decision tree with adaboost).

- ◎ Travel time of every 30 minutes
- ◎ Travel time of every hour

Decision Tree Regression with Adaboost

Feature

- Start point time
- Exit path
- Average travel time of last 30 minutes
- Average travel time of last one hour

Target

- Total travel time

Result

max_depth=5; max_depth=5

- 0.2058

max_depth=5; max_depth=50

- 0.2049

max_depth=20; max_depth=20

- 0.2045

According to the previous result, we thought that if using max_depth=5, the result might be the best. However, the real result does not match what we thought.

Problem

Noise correction is not complete.

There exist null values, but the total travel time is inside the range of 25%-75%.

4.55	5.45	13.2	0	0	7.66	A	2016/7/25 07:55	3	201.28
3.86	4	10.3	0	0	5.41	A	2016/7/25 07:55	3	175.33
3.7	4.75	13.89	117.21	0.86	5.8	A	2016/7/25 07:57	2	146.21
3.18	3.89	13.55	0	0	6.18	A	2016/7/25 08:02	3	179.41
5.23	6.15	37.81	0	0	15.25	A	2016/7/25 08:07	3	157.31
3.63	4.27	13.17	0	0	6.3	A	2016/7/25 08:08	3	191.41
5.22	6.29	13.44	93.69	0.97	6.72	A	2016/7/25 08:09	2	126.69

Thanks!