



EXPLORE ADVERSE EVENTS ASSOCIATED WITH CONCOMITANT MEDICATIONS USING PREDICTION MODEL ESTABLISHED ON GRADIENT BOOSTED DECISION TREE (GBDT) METHOD

Yang Chong
Senior Program and Application Developer
Merck Serono



Data Mining

Data Mining is the process of discovering patterns in large data sets involving methods at the intersection of machine learning, statistics, and database systems.

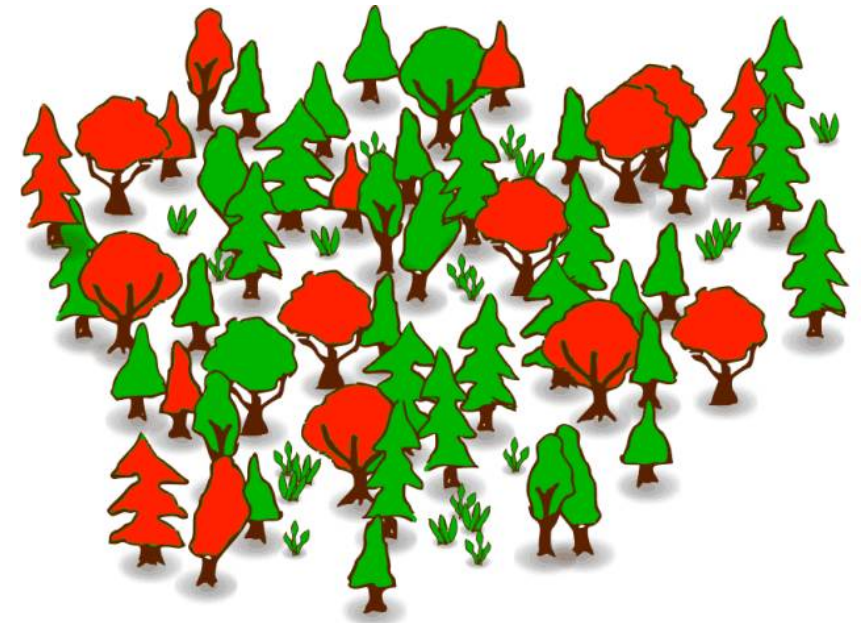
Data Mining is the analysis step of the "knowledge discovery in databases" process, or KDD.

***"We're drowning in
information, but starving for
knowledge."***

Machine Learning

Machine Learning is a field of computer science that uses statistical techniques to give computer systems the ability to "learn" (i.e., progressively improve performance on a specific task) with data, without being explicitly programmed.

- Machine Learning explores the study and construction of algorithms that can learn from and make predictions on data.
- Machine Learning is the foundation of Artificial Intelligence.



Gradient Boosting Decision Tree (GBDT)

Input: training set $\{(x_i, y_i)\}_{i=1}^n$, a differentiable loss function $L(y, F(x))$, number of iterations M .

Algorithm:

1. Initialize model with a constant value:

$$F_0(x) = \arg \min_{\gamma} \sum_{i=1}^n L(y_i, \gamma).$$

2. For $m = 1$ to M :

1. Compute so-called *pseudo-residuals*:

$$r_{im} = - \left[\frac{\partial L(y_i, F(x_i))}{\partial F(x_i)} \right]_{F(x)=F_{m-1}(x)} \quad \text{for } i = 1, \dots, n.$$

2. Fit a base learner (e.g. tree) $h_m(x)$ to pseudo-residuals, i.e. train it using the training set $\{(x_i, r_{im})\}_{i=1}^n$.

3. Compute multiplier γ_m by solving the following one-dimensional optimization problem:

$$\gamma_m = \arg \min_{\gamma} \sum_{i=1}^n L(y_i, F_{m-1}(x_i) + \gamma h_m(x_i)).$$

4. Update the model:

$$F_m(x) = F_{m-1}(x) + \gamma_m h_m(x).$$

3. Output $F_M(x)$.

Data Mining in Clinical Trial

Data Sources

- Historical Trials
- Electronic Health Record
- Life Style, Vital Signs
- Drug database
- Hospital Data
- Patent/Scientific Publication
- Social Media

Machine Learning

- Association Rules
- Anomaly Detection
- Clustering
- Classification
- Prediction
- Natural Language Processing

Objectives

- Drug Discovering
- Time & Cost Reducing
- Site & Subject Selection
- Quality Control
- Adverse Event alert
- Query Prediction
- Cycle Time Risk
- Efficacy & Safety Assessment
- Visualization

Concomitant Medications

CMTRT	CMDECOD	CMINDC	CMDOSE	CMDOSU	CMSTDTC	CMENDTC	CMSTDY	CMENDY	CMINDAE
MANNITOL		HYDRATION	125 mL		2016-01-08	2016-01-08	2	2	
PALONOSETRON	PALONOSETRON	PREVENTING V...	0.5 mg		2016-01-08	2016-01-12	2	6	
PANTOLOC		INHIBITS GAST...	40 mg		2016-01-08	2016-01-09	2	3	
LACTULOSE	LACTULOSE	RELAXING THE ...	15 mL		2016-01-09	2016-01-09	3	3	Y
CETYLPYRIDINI...		PREVENTION O...	15 mL		2016-01-10	2016-01-10	4	4	
MEGESTROL	MEGESTROL	PREVENT LOW ...	160 mg		2016-01-10	2016-01-10	4	4	
MOTILUM	MOTILUM ...	IMPROVE THE ...	10 mg		2016-01-10	2016-01-10	4	4	
MEGESTROLAC...	MEGESTROL ACE...	IMPROVES APP...	160 mg		2016-01-13	2016-01-14	7	8	Y
CARBOHYDRATE	CARBOHYDRATES	NUTRITION SU...	55.8 g		2016-01-14	2016-01	8	.	
CETYLPYRIDINI...		PREVENTION O...	15 mL		2016-01-14	2016-01-14	8	8	Y
KANGFUXIN LO...		PREVENTION O...	25 mL		2016-01-14	2016-01-14	8	8	Y
LACTULOSE	LACTULOSE	RELAXING THE ...	15 mL		2016-01-20	2016-01-20	14	14	Y
CETYLPYRIDINI...		PREVENT ORAL...	15 mL		2016-01-21	2016-01-21	15	15	Y
KANGFUXING L...	UNSPECIFIED HE...	PREVENT ORAL...	25 mL		2016-01-21	2016-01-21	15	15	Y

Adverse Events

AETERM	AEDECOD	AEPTCD	AESOC	AESOCCD	AESER	AEREL	AETOXGR	AESTDTC	AEENDTC	AESTDY	AEENDY
CHEST DISTR...	Chest discomfort	10008469	General disorders...	10018065		UNRELATED	1	2016-01-04	2016-01-06	-3	-1
CONSTIPATION	Constipation	10010774	Gastrointestinal di...	10017947		UNRELATED	2	2016-01-05	2016-01-11	-2	5
CONSTIPATION	Constipation	10010774	Gastrointestinal di...	10017947		UNRELATED	1	2016-01-20	2016-01-21	14	15
CONSTIPATION	Constipation	10010774	Gastrointestinal di...	10017947		UNRELATED	1	2016-03-07	2016-03-08	61	62
DIZZY	Dizziness	10013573	Nervous system ...	10029205		UNRELATED	1	2016-02-14	2016-02-15	39	40
DIZZY	Dizziness	10013573	Nervous system ...	10029205		UNRELATED	1	2016-04-19	2016-04-20	104	105
HYPOALBUMIN...	Hypoalbuminaemia	10020942	Metabolism and n...	10027433		UNRELATED	1	2016-01-05	2016-01-27	-2	21
HYPOALBUMIN...	Hypoalbuminaemia	10020942	Metabolism and n...	10027433		UNRELATED	1	2016-02-03	2016-02-14	28	39
HYPOALBUMIN...	Hypoalbuminaemia	10020942	Metabolism and n...	10027433		UNRELATED	1	2016-02-23	2016-03-11	48	65
HYPOALBUMIN...	Hypoalbuminaemia	10020942	Metabolism and n...	10027433		UNRELATED	1	2016-03-21	2016-04-05	75	90
HYPOALBUMIN...	Hypoalbuminaemia	10020942	Metabolism and n...	10027433		UNRELATED	1	2016-04-13		98	.
HYPOCHLORID...	Hypochloraemia	10020955	Metabolism and n...	10027433		UNRELATED	1	2016-01-14	2016-01-21	8	15
HYPOCHLORID...	Hypochloraemia	10020955	Metabolism and n...	10027433		UNRELATED	1	2016-04-13		98	.
HYPOKALEMIA	Hypokalaemia	10021015	Metabolism and n...	10027433		UNRELATED	1	2016-03-21	2016-04-05	75	90
HYPOKALEMIA	Hypokalaemia	10021015	Metabolism and n...	10027433		UNRELATED	1	2016-04-13		98	.

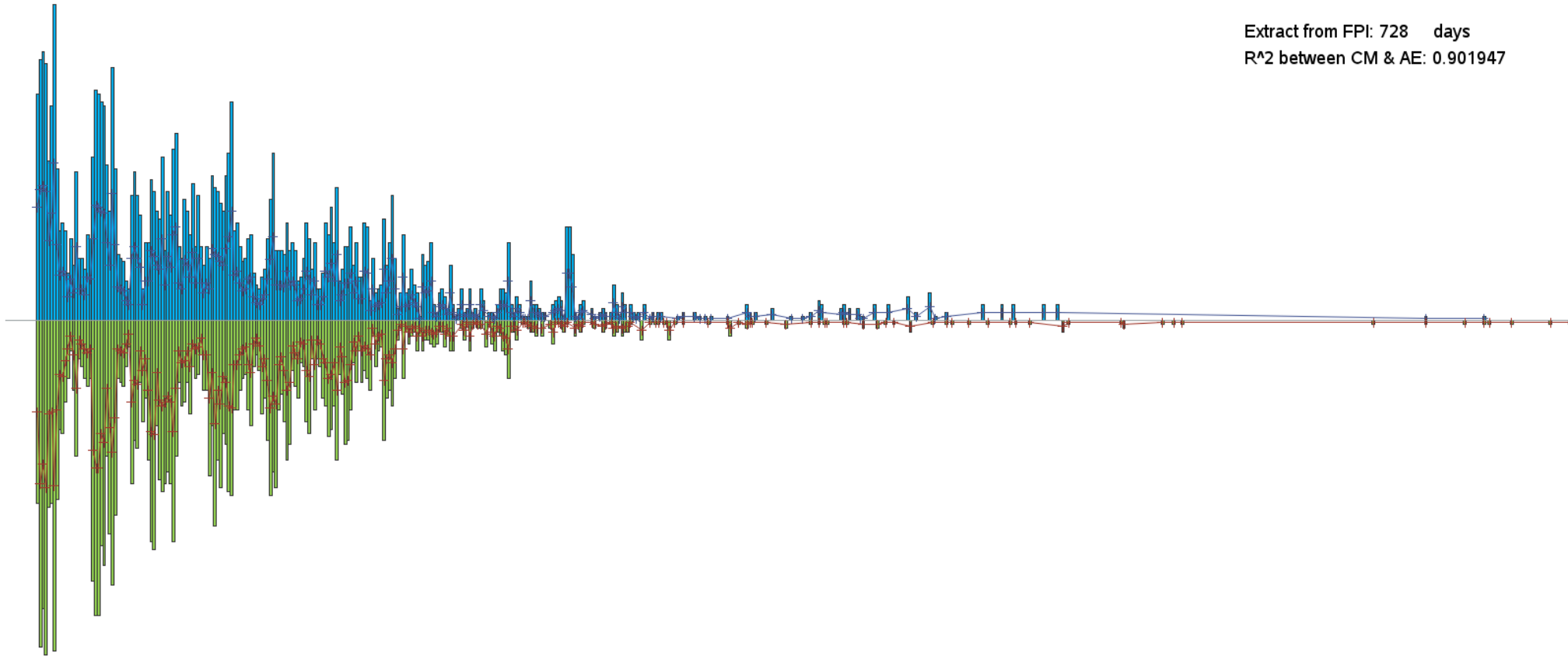
The Relationship Between CM & AE

Relationship Between CM And AE

CM n = XXXX

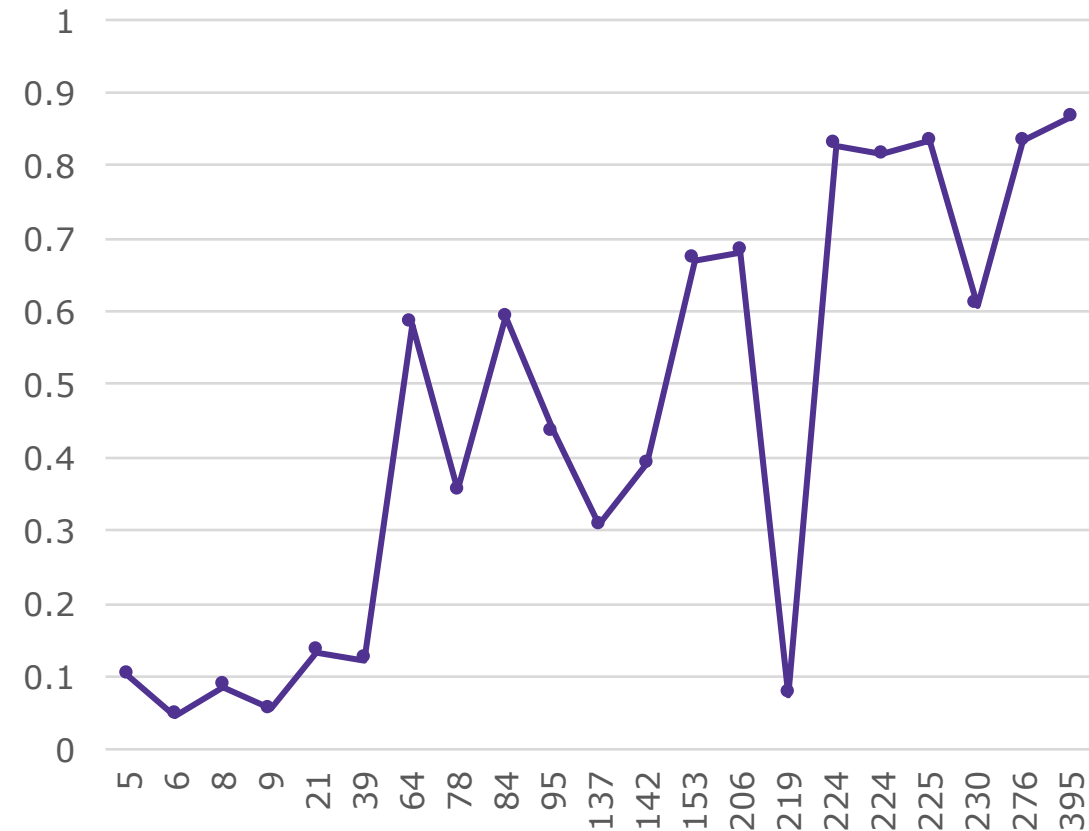
AE n = XXXX

Extract from FPI: 728 days
R² between CM & AE: 0.901947



The Relationship Between R^2 (CM vs AE) & Study Days

study	n	r^2
Merck001	8	0.085714
Merck002	219	0.074881
Merck003	64	0.583836
Merck004	142	0.390423
Merck005	230	0.60832
Merck006	395	0.867134
Merck007	276	0.833937
Merck008	206	0.682628
Merck009	224	0.827393
Merck010	9	0.054054
Merck011	153	0.669992
Merck012	39	0.121783
Merck013	95	0.434316
Merck014	137	0.306424
Merck015	5	0.10101
Merck016	224	0.815517
Merck017	6	0.045455
Merck018	21	0.132726
Merck019	78	0.35424
Merck020	84	0.592257
Merck021	225	0.833028



Data Space

Expanding CMDECOD (n = 2245) & AEDECOD (n = 2370) by subject by event start date from Day 1 (n = 7858)

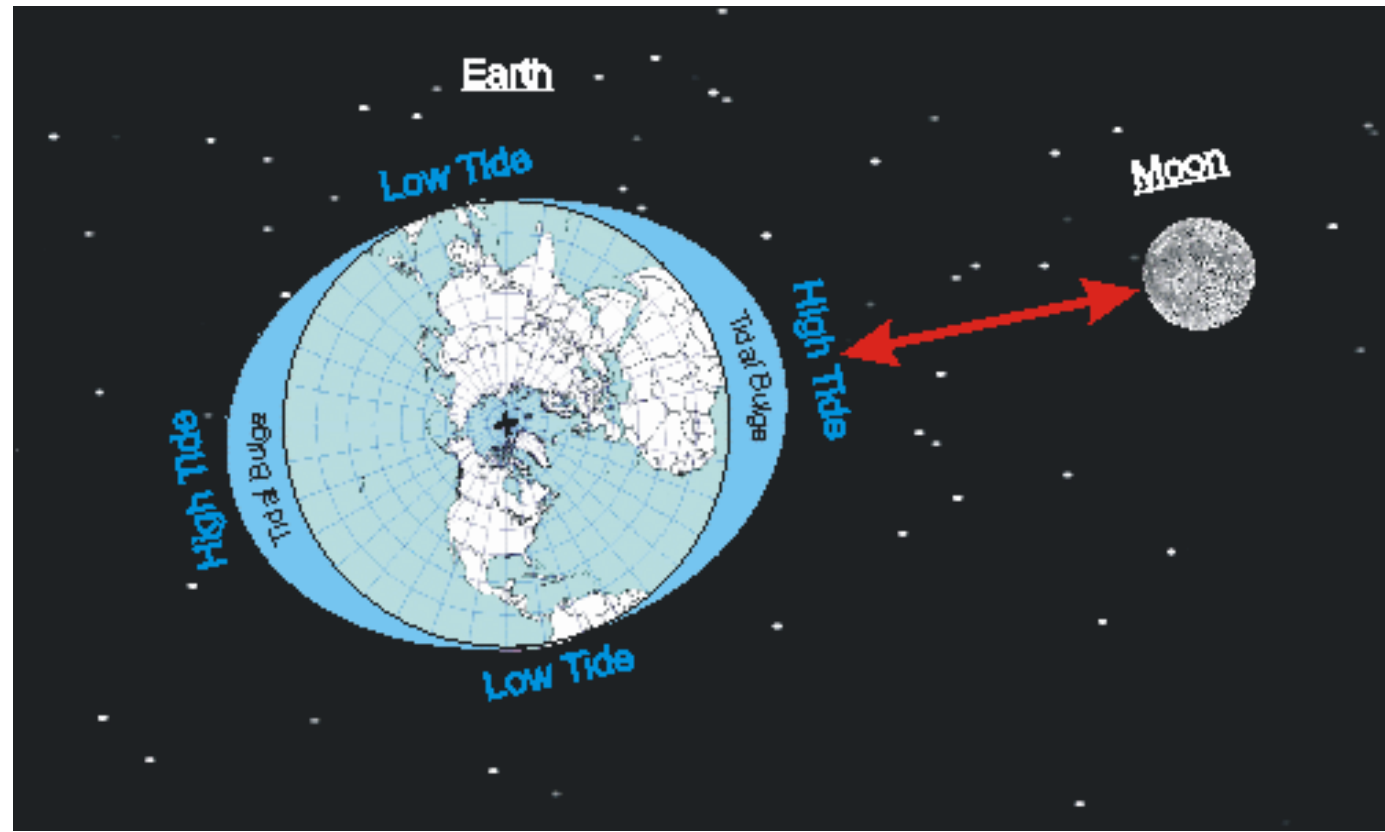
stdy	CM_FU ROSEM IDE	CM_CO NTAC	CM_FL UTICAS ONE	CM_DU ONASE	CM_LO PERAMI DE_HY DROCH LORIDE	CM_ON DANSE TRON	CM_EP OPROS TENOL SODIU M	CM_IBU PROFE N	CM_AZI THRO MYCIN	CM_PR OMETH AZINE_ W_COD EINE	CM_BE NZONA TATE	CM_LE VOFLO XACIN	CM_PR EDNIS ONE	CM_SO DIUM_ CHLORI DE	AE_Abd ominal_ discomf ort	AE_Hyp ertensio n	AE_Pul monary_ hypert ension	AE_Fati gue	AE_Diar rhea	AE_Epis taxis	AE_Skin _ulcer	AE_Dys pnoea	AE_Dizz iness	AE_Hea dache	AE_Nau sea	AE_Vo miting
141	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	0	0	0	0	0	0	0	0	0
157	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
185	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
98	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
120	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0
15	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
18	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
218	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
259	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
235	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
43	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
159	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
247	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
255	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
143	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
214	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
-10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
79	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0

decod_stdtc_train_01

decod_stdtc_train_01

Parameters Setting in PYTHON

```
gbdt=GradientBoostingRegressor(  
    loss='ls'  
    , learning_rate=0.1  
    , n_estimators=100  
    , subsample=1  
    , min_samples_split=2  
    , min_samples_leaf=1  
    , max_depth=3  
    , init=None  
    , random_state=None  
    , max_features=None  
    , alpha=0.9  
    , verbose=0  
    , max_leaf_nodes=None  
    , warm_start=False)
```



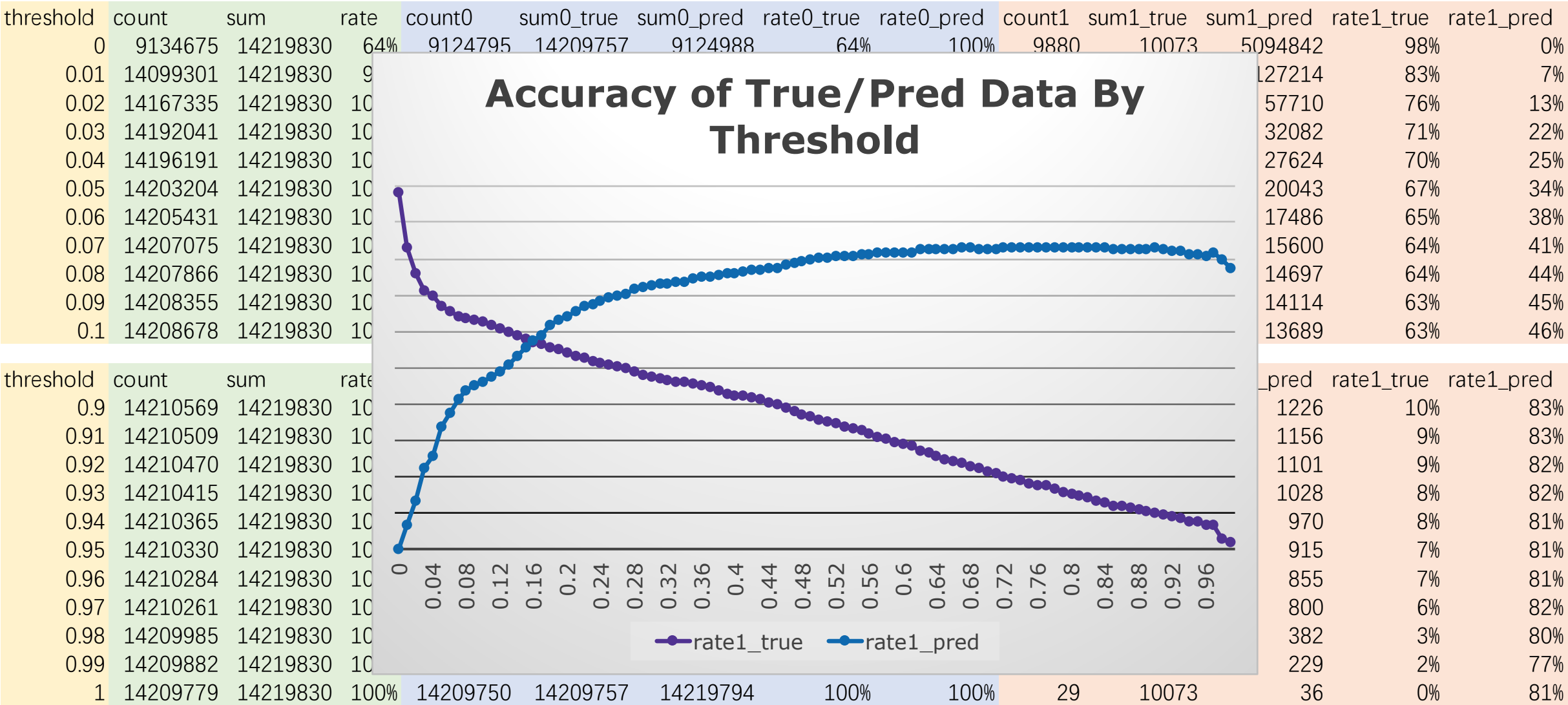
Predicted AE Values

Predicted AE values are floating-point number

A threshold is needed for 0-1 classification

AE_Abdon	AE_Hypert	AE_Pulmor	AE_Fatigue	AE_Diarrhc	AE_Epistax	AE_Skin_uli	AE_Dyspnc	AE_Dizzine	AE_Headac	AE_Nausea	AE_Vomitir	AE_Arthral	AE_Hypoa	AE_Muscul	AE_Neck_p	AE_Product	AE_Viral_Ur
-6.02E-05	0.006246	0.972191	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.00208	0.002005
-6.02E-05	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.841245	0.774686
-0.00017	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.008017	0.057116
-0.00017	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.00208	0.002005
0.000151	0.006246	3.54E-06	0.016553	0.838284	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.00208	0.002005
0.000109	0.006246	3.54E-06	0.016553	0.015887	0.000298	0.97737	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.02224	0.015887	0.000298	4.23E-05	0.015821	0.006379	0.006636	0.026376	0.017067	0.005317	0.000316	0.001259	0.001404	0.00208	0.002005
0.009624	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.176195	0.022939	0.017067	0.037698	0.002828	0.000823	0.029824	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.070003	0.0032	0.006636	0.022939	0.017067	0.005317	-8.57E-05	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-8.57E-05	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-8.57E-05	0.000823	0.001404	0.00208	0.002005
0.000151	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.133579	0.022939	0.017067	0.010352	-0.00014	0.000823	0.001404	0.00208	0.01802
0.000193	0.006246	3.54E-06	0.020278	0.015887	0.000298	4.45E-05	0.014303	0.0032	0.006636	0.028207	0.017067	0.005317	-0.00032	0.006311	0.001404	0.00208	0.002005
6.51E-05	0.61432	3.54E-06	0.020278	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.028207	0.017067	0.005317	-0.00032	0.006311	0.001404	0.00208	0.002005
0.009624	0.006246	3.54E-06	0.020278	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.120681	0.026376	0.017067	0.037698	0.006009	0.001259	0.029824	0.00208	0.002005
0.009624	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.120681	0.022939	0.017067	0.037698	0.002879	0.000823	0.029824	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-8.57E-05	0.031553	0.001404	0.00188	0.002005
-1.08E-06	0.006246	3.54E-06	0.020278	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.026376	0.017067	0.005317	0.000316	0.0015	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.117079	0.022939	0.017067	0.012619	0.001988	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.017426	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.100662	-8.57E-05	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-0.00014	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.020278	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.026376	0.017067	0.005317	-8.57E-05	0.000823	0.001404	0.00208	0.002005
-1.08E-06	0.006246	3.54E-06	0.016553	0.015887	0.000298	4.45E-05	0.015821	0.0032	0.006636	0.022939	0.017067	0.005317	-8.57E-05	0.000823	0.001404	0.00208	0.002005

Threshold Setting



Scenario: AE risk

Threshold = 1:

		True AE		
		Yes	No	Total
Pred AE	Yes	29	7	36
	No	10044	14209750	14219794
	Total	10073	14209757	14219830

Sensitivity = $29/10073 = 0.29\%$

Specificity = $14209750/14209757 = 99.99\%$

Accuracy of the model = $(29 + 14209790)/14219830 = 99.93\%$

False positive rate = $7/14209757 = 0\%$

False negative rate = $10044/10073 = 99.71\%$

subject006 2 AE_Vomiting

CMTRT	CMDECOD	CMSTDTC	CMSTDY ▲
20% MANNITOL	MANNITOL	2015-10-30	2
METOCLOPRAM...	METOCLOPRAMIDE DIHYDROCHLORIDE	2015-10-30	2
PIDOTIMOD OR...	PIDOTIMOD	2015-10-30	2

AETERM	AEDECOD	AESTDTC	AESTDY ▲
NAUSEA	Nausea	2015-10-30	2

AETERM	AEDECOD	AESTDTC	AESTDY ▲
NAUSEA	Nausea	2015-10-30	2
VOMIT	Vomiting	2015-10-31	3
VOMIT	Vomiting	2015-10-31	3



Q & A