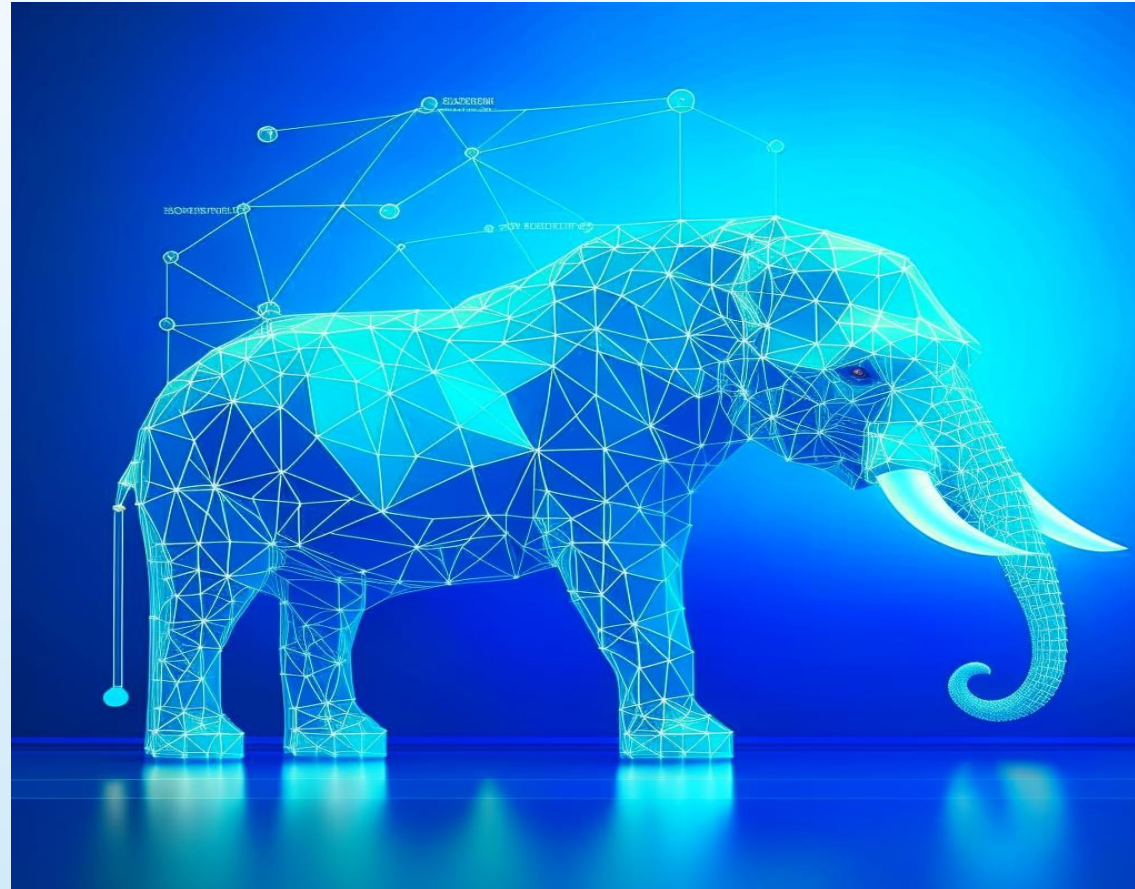


ML in PostgreSQL



Using models:

- Binary Classification
- Multi Classification
- Regression

Using framework:

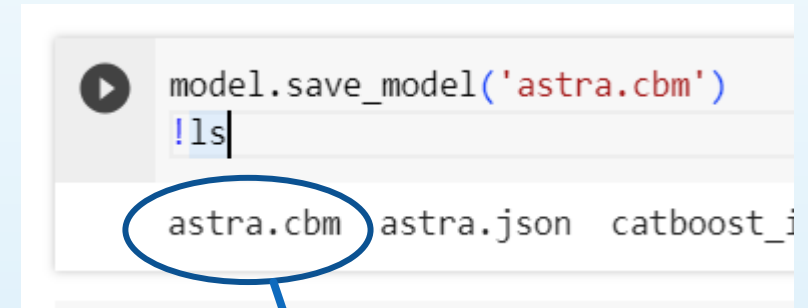


CatBoost: <https://catboost.ai/>

only prediction

ML process:

1. Training model
2. Save model to database server



3. prediction

```
adult=# SELECT ml_predict('astra3.cbm', 'astra3');  
WARNING: field run_id not used  
WARNING: field field_id not used  
WARNING: field spec_obj_id not used  
WARNING: field predict not used  
      ml_predict  
-----  
public.astra3_predict  
(1 row)
```

Installation

- ❑ git clone https://github.com/akalend/pg_ml.git
- ❑ export PG_HOME=/usr/local/pgsql //where is main postgres folder
- ❑ wget <https://github.com/catboost/catboost/releases/download/v1.2.2/libcatboostmodel.so>
- ❑ mv libcatboostmodel.so \$PG_HOME/lib
- ❑ cd pg_ml
- ❑ export PG_CONFIG=\$PG_HOME/bin/pg_config
- ❑ export LD_LIBRARY_PATH=\$PG_HOME/lib
- ❑ USE_PGXS=1 make
- ❑ sudo su
- ❑ export PATH=\$PATH:\$PG_HOME/bin
- ❑ USE_PGXS=1 make install
- ❑ chown postgres model.cbm
- ❑ [optional] cp model.cbm \$PG_HOME/data



Configuration

postgresql.conf:

path to model folder

ml.model_path = /usr/local/model

DataFrame columns convert to PostgreSQL fields



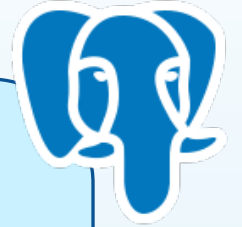
ID, Id, id

FieldID, Field_ID

Import-port

Port, PORT

<=>



id

field_id

import_port

port

DataFrame columns and PostgreSQL fields

```
[14] df = pd.read_csv('star_classification.csv')
```

```
[16] for it in df.columns:  
      print(it)
```

```
obj_ID  
alpha  
delta  
u  
g  
r  
i  
z  
run_ID  
rerun_ID  
cam_col  
field_ID  
spec_obj_ID  
class  
redshift  
plate  
MJD  
fiber_ID
```



```
adult=# \d astra3
```

Column	Type	Collation
alpha	double precision	
delta	double precision	
u	double precision	
g	double precision	
r	double precision	
i	double precision	
z	double precision	
run_id	bigint	
cam_col	bigint	
field_id	bigint	
spec_obj_id	double precision	
redshift	double precision	
plate	bigint	
mjd	bigint	
fiber_id	bigint	

```
adult=#
```



DataFrame columns and PostgreSQL fields



```
df.head()
```

	obj_ID	alpha	delta	u	g	r	i	z	run_ID	rerun_ID	cam_col	field_ID	spec_obj_ID
0	1.237661e+18	135.689107	32.494632	23.87882	22.27530	20.39501	19.16573	18.79371	3606	301	2	79	6.543777e+18
1	1.237665e+18	144.826101	31.274185	24.77759	22.83188	22.58444	21.16812	21.61427	4518	301	5	119	1.176014e+19
2	1.237661e+18	142.188790	35.582444	25.26307	22.66389	20.60976	19.34857	18.94827	3606	301	2	120	5.152200e+18
3	1.237663e+18	338.741038	-0.402828	22.13682	23.77656	21.61162	20.50454	19.25010	4192	301	3	214	1.030107e+19
4	1.237680e+18	345.282593	21.183866	19.43718	17.58028	16.49747	15.97711	15.54461	8102	301	3	137	6.891865e+18

```
adult=# select * from astra3 limit 3;
```

alpha	delta	u	g	r	i	z	run_id	cam_col	field_id	spec_obj_id
16.9568897845004	3.64613008870454	23.33542	21.95143	20.48149	19.603	19.13094	7712	6	442	4.855016555329904e+18
240.063240247767	6.13413059813973	17.86033	16.79228	16.43001	16.30923	16.25873	3894	1	243	2.4489280322708705e+18
30.887222067625	1.18870964120799	18.18911	16.89469	16.42161	16.24627	16.18549	7717	1	536	8.255357438959835e+18

```
(3 rows)
```



Information about model

```
adult=# SELECT ml_info ('astra3.cbm');
          ml_info
-----
dimension:3 numeric features:12 categorical features:0 modelType "MultiClass"+
fieldName:alpha,delta,u,g,r,i,z,can_col,redshift,plate,MJD,fiber_ID
(1 row)
adult=#
```

- Dimension result (How many classes)
- Feature count (categorical and float)
- Type of model
- Fields name

Information about model

```
adult=# SELECT ml_info ('astra3.cbm');
          ml_info
-----
dimension:3 numeric features:12 categorical features:0 modelType "MultiClass"+
fieldName:alpha,delta,u,g,r,i,z,can_col,redshift,plate,MJD,fiber_ID
(1 row)

adult=#
```

```
adult=# SELECT ml_info ('titanic.cbm');
          ml_info
-----
dimension:1 numeric features:2 categorical features:9 modelType "Accuracy" +
fieldName:PassengerId,Pclass,Name,Sex,Age,SibSp,Parch,Ticket,Fare,Cabin,Embarked
(1 row)
```

Model information

```
akalend@notebook-sasha: ~/stars
adult=# select ml_info('boston.cbm');
          ml_info
-----
dimension:1 numeric features:13 categorical features:0 modelType "RMSE" +
fieldName:crim,zn,indus,chas,nox,rm,age,dis,rad,tax,ptratio,black,lstat
(1 row)
adult=#
```

More information

```
model.save_model('astra.json', format='json')  
!ls  
astra.cbm astra.json catboost_info sample_data
```

```
adult=# SELECT * from ml_json_info('astra.json');  
ml_json_info  
-----  
float feature:alpha,delta,u,g,r,i,z,cam_col,redshift,plate,MJD,fiber_ID,+  
categorical feature:  
(1 row)
```

Categorical Feature list
Float Feature count list

Prediction of model (recordset)

Path to model file

Table name

Categorical field list

```
adult=# SELECT * from ml_cat_predict ('titanic.cbm',  
'titanic', '{name,passenger_id,pclass,sex,sibsp,parch,ticket,cabin,embarked }');
```

row_num	predict	class
0	-1.7937342449233795	0
1	-0.7958399022225136	0
2	-2.392873216013247	0
3	-1.942976624899004	0
4	-0.41747860726736713	0
5	-2.0608914711097546	0
6	0.5914467057444344	1
7	-1.0786526230973736	0
8	0.6757411102494171	1
9	-3.250956928980716	0
10	-2.274725588104562	0
11	-1.3228896775643357	0
12	2.70931909246417	1
13	-2.4233542239140187	0

Prediction of model (query)

Path to model file

query

```
postgres@notebook-sasha: /usr/local/pgsql
postgres@notebook-sasha:/usr/local/pgsql$ bin/psql adult
psql (15.1)
Type "help" for help.

adult=# SELECT * FROM ml_predict_query ('astra3.cbm', 'SELECT * FROM astra3');
 index |          predict          | class
-----+-----+-----
  0    | 0.9868595777513302       | GALAXY
  1    | 0.9904188657285139       | STAR
  2    | 0.9975875623929414       | STAR
  3    | 0.9976669380943318       | STAR
  4    | 0.9960439244920889       | STAR
  5    | 0.9843734017027631       | GALAXY
  6    | 0.9961635567874662       | GALAXY
  7    | 0.9577871819302538       | QSO
  8    | 0.9938922568658763       | GALAXY
  9    | 0.9938564131331261       | GALAXY
 10    | 0.9988845094177173       | GALAXY
 11    | 0.993620291192055        | GALAXY
 12    | 0.9864315757824991       | GALAXY
 13    | 0.9990696704299644       | STAR
 14    | 0.9678149544851397       | GALAXY
 15    | 0.9834028383120225       | GALAXY
 16    | 0.9982105605728178       | STAR
 17    | 0.9947704704037497       | GALAXY
 18    | 0.9948908257380777       | GALAXY
 19    | 0.9937357203427619       | GALAXY
 20    | 0.996724214136651        | GALAXY
 21    | 0.9981412663503989       | GALAXY
```

Prediction of model (result table)

Path to model file

Table name

```
adult=# SELECT * from ml_predict_table('astra3.cbm','astra3');
ml_predict_table
-----
public.astra3_predict (1 row)
```

**Create
the new
table**

Prediction of model (table)

Path to model file

Table name

List of categorical fields

```
adult=# SELECT ml_predict ('adult.cbm', 'adult2',  
adult(# '{workclass,education,marital_status, occupation,relationship,race,sex,native_country}');  
ml_predict  
-----  
public.adult2_predict  
(1 row)
```

**Create
the new
table**

Prediction results

SELECT * FROM {table}_predict;

adult=# SELECT * from astra3_predict;																	
row	alpha	delta	u	g	r	i	z	run_id	cam_col	field_id	spec_obj_id	redshift	plate	mjd	fiber_id	predict	class
1	16.9568897845004	3.64613008870454	23.33542	21.95143	20.48149	19.603	19.13094	7712	6	442	4.855016555329904e+18	0.5062369	4312	55511	495	0.98686	GALAXY
2	240.063240247767	6.13413059813973	17.86033	16.79228	16.43001	16.30923	16.25873	3894	1	243	2.4489280322708705e+18	0.0003448142	2175	54612	348	0.990419	STAR
3	30.887222067625	1.18870964120799	18.18911	16.89469	16.42161	16.24627	16.18549	7717	1	536	8.255357438959835e+18	4.085216e-06	7332	56683	943	0.997588	STAR
4	247.594400505002	10.8877797153666	24.99961	21.71203	21.47148	21.30532	21.29109	5323	1	134	4.577998722756271e+18	-0.0002914838	4066	55444	326	0.997667	STAR
5	18.8964507920807	-5.26133022886992	23.76648	21.79737	20.69543	20.23403	19.97464	7881	3	148	8.91047176642785e+18	-0.0001361561	7914	57331	363	0.996044	STAR
6	182.713733094955	51.3758050594777	22.44608	21.68444	20.24292	19.41423	19.08227	2830	1	411	7.516725588574623e+18	0.5026683	6676	56389	792	0.984373	GALAXY
7	150.089423193165	39.4670880748061	18.96441	17.82906	17.31429	16.99891	16.85583	3560	4	278	1.5267956411104236e+18	0.06366445	1356	53033	274	0.996164	GALAXY
8	189.510984338851	58.7411197772507	21.37376	20.80187	20.84925	21.13449	20.34689	2243	1	353	7.696817897528907e+18	0.7936153	6836	56443	604	0.957787	QSO
9	37.7138728560977	-0.525138228146508	20.77988	19.54618	19.16687	18.89438	18.64286	2700	2	117	1.7553283123029217e+18	0.1060118	1559	53271	183	0.993892	GALAXY
10	201.074980072746	28.7699058867715	25.05349	22.23362	20.8122	19.69488	19.28336	4649	3	120	7.306035245308205e+18	0.567082	6489	56329	257	0.993856	GALAXY
11	151.83091832672	19.8108624669417	24.04443	22.48608	20.59701	19.50985	19.00457	5183	5	142	6.622787444780849e+18	0.5475619	5882	56029	888	0.998885	GALAXY

SELECT * FROM
ml_predict(...);

adult=# SELECT * from ml_predict('astra3.cbm' 'astra3');		
id	predict	class
0	0.9868595777513302	GALAXY
1	0.9904188657285139	STAR
2	0.9975875623929414	STAR
3	0.9976669380943318	STAR
4	0.9960439244920889	STAR
5	0.9843734017027631	GALAXY
6	0.9961635567874662	GALAXY
7	0.9577871819302538	QSO
8	0.9938922568658763	GALAXY
9	0.9938564131331261	GALAXY

Binary classification

postgres@notebook-sasha: /usr/local/pgsql

```
adult=# select * from titanic_predict;
```

row	id	passenger_id	pclass	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	res	predict	clas
1	0	892	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	-999	Q	f	0.142616	0
2	1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47	1	0	363272	7	-999	S	f	0.310916	0
3	2	894	2	Myles, Mr. Thomas Francis	male	62	0	0	240276	9.6875	-999	Q	f	0.083718	0
4	3	895	3	Wirz, Mr. Albert	male	27	0	0	315154	8.6625	-999	S	f	0.125321	0
5	4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22	1	1	3101298	12.2875	-999	S	f	0.39712	0
6	5	897	3	Svensson, Mr. Johan Cervin	male	14	0	0	7538	9.225	-999	S	f	0.112956	0
7	6	898	3	Connolly, Miss. Kate	female	30	0	0	330972	7.6292	-999	Q	t	0.643697	1
8	7	899	2	Caldwell, Mr. Albert Francis	male	26	1	1	248738	29	-999	S	f	0.253761	0
9	8	900	3	Abraham, Mrs. Joseph (Sophie Halaut Easu)	female	18	0	0	2657	7.2292	-999	C	t	0.662787	1
10	9	901	3	Davies, Mr. John Samuel	male	21	2	0	A/4 48871	24.15	-999	S	f	0.037293	0
11	10	902	3	Iliffe, Mr. Yllo	male	-999	0	0	349220	7.8958	-999	S	f	0.093238	0
12	11	903	1	Jones, Mr. Charles Cresson	male	46	0	0	694	26	-999	S	f	0.210338	0

```
adult=# SELECT * from ml_cat_predict ('titanic.cbm', 'titanic', '{name,passenger_id,pclass,sex,sibsp,parch,ticket,cabin,embarked}');
```

row_num	predict	class
0	-1.7937342449233795	0
1	-0.7958399022225136	0
2	-2.392873216013247	0
3	-1.942976624899004	0
4	-0.41747860726736713	0
5	-2.0608914711097546	0
6	0.5914467057444344	1
7	-1.9769579739773739	0

Binary classification

```
adult=# SELECT * FROM ml_cat_predict ('adult.cbm', 'adult2','{workclass,
education,marital_status, occupation,relationship,race,sex,native_country}
');
```

row_num	predict	class
0	-5.926338548423682	<=50K
1	-1.225876230403332	<=50K
2	-0.7485117670534811	<=50K
3	3.6351647093731705	>50K
4	-4.644606242153101	<=50K
5	-5.342578732065899	<=50K
6	-3.9224526779262296	<=50K

postgres@notebook-sasha: /usr/local/pgsql																
row	age	workclass	fnlwgt	education	education_num	marital_status	occupation	relationship	race	sex	capital_gain	capital_loss	hours_per_week	native_country	predict	class
1	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	0	40	United-States	0.002661	<=50K
2	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	0	50	United-States	0.226904	<=50K
3	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	0	40	United-States	0.321146	<=50K
4	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	0	40	United-States	0.974298	>50K
5	18	nan	103497	Some-college	10	Never-married	nan	Own-child	White	Female	0	0	30	United-States	0.009522	<=50K
6	34	Private	198693	10th	6	Never-married	Other-service	Not-in-family	White	Male	0	0	30	United-States	0.004761	<=50K
7	29	nan	227026	HS-grad	9	Never-married	nan	Unmarried	Black	Male	0	0	40	United-States	0.019408	<=50K
8	63	Self-emp-not-inc	104626	Prof-school	15	Married-civ-spouse	Prof-specialty	Husband	White	Male	3103	0	32	United-States	0.560734	>50K
9	24	Private	369667	Some-college	10	Never-married	Other-service	Unmarried	White	Female	0	0	40	United-States	0.003572	<=50K
10	55	Private	104996	7th-8th	4	Married-civ-spouse	Craft-repair	Husband	White	Male	0	0	10	United-States	0.092036	<=50K
11	65	Private	104454	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	6418	0	40	United-States	0.759614	>50K
12	36	Federal-gov	212465	Bachelors	13	Married-civ-spouse	Adm-clerical	Husband	White	Male	0	0	40	United-States	0.64466	>50K
13	26	Private	82091	HS-grad	9	Never-married	Adm-clerical	Not-in-family	White	Female	0	0	39	United-States	0.007208	<=50K
14	58	nan	299831	HS-grad	9	Married-civ-spouse	nan	Husband	White	Male	0	0	35	United-States	0.432042	<=50K

--Далее--

Regression

akalend@notebook-sasha: ~/stars

row	index	crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	black	lstat	medv	predict
1	0	0.00632	18	2.31	0	0.538	6.575	65.2	4.09	1	296	15.3	396.9	4.98	24	24.99982
2	1	0.02731	0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.9	9.14	21.6	20.664359
3	2	0.02729	0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83	4.03	34.7	33.677379
4	3	0.03237	0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63	2.94	33.4	34.289002
5	4	0.06905	0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.9	5.33	36.2	34.615708
6	5	0.02985	0	2.18	0	0.458	6.43	58.7	6.0622	3	222	18.7	394.12	5.21	28.7	27.968317
7	6	0.08829	12.5	7.87	0	0.524	6.012	66.6	5.5605	5	311	15.2	395.6	12.43	22.9	21.682186
8	7	0.14455	12.5	7.87	0	0.524	6.172	96.1	5.9505	5	311	15.2	396.9	19.15	27.1	22.853984
9	8	0.21124	12.5	7.87	0	0.524	5.631	100	6.0821	5	311	15.2	386.63	29.93	16.5	17.011092
10	9	0.17004	12.5	7.87	0	0.524	6.004	85.9	6.5921	5	311	15.2	386.71	17.1	18.9	18.24062
11	10	0.22489	12.5	7.87	0	0.524	6.377	94.3	6.3467	5	311	15.2	392.52	20.45	15	17.543837
12	11	0.11747	12.5	7.87	0	0.524	6.009	82.9	6.2267	5	311	15.2	396.9	13.27	18.9	19.974764
13	12	0.09378	12.5	7.87	0	0.524	5.889	39	5.4509	5	311	15.2	390.5	15.71	21.7	20.70866
14	13	0.62976	0	8.14	0	0.538	5.949	61.8	4.7075	4	307	21	396.9	8.26	20.4	20.202922
15	14	0.63796	0	8.14	0	0.538	6.096	84.5	4.4619	4	307	21	380.02	10.26	18.2	18.175456
16	15	0.62739	0	8.14	0	0.538	5.834	56.5	4.4986	4	307	21	395.62	8.47	19.9	19.698336
17	16	1.05393	0	8.14	0	0.538	5.935	29.3	4.4986	4	307	21	386.85	6.58	23.1	22.304514
18	17	0.7842	0	8.14	0	0.538	5.99	81.7	4.2579	4	307	21	386.75	14.67	17.5	17.160698
19	18	0.80271	0	8.14	0	0.538	5.456	36.6	3.7965	4	307	21	288.99	11.69	20.2	18.706903
20	19	0.7258	0	8.14	0	0.538	5.727	69.5	3.7965	4	307	21	390.95	11.28	18.2	18.769804
21	20	1.25179	0	8.14	0	0.538	5.57	98.1	3.7979	4	307	21	376.57	21.02	13.6	13.985032

--Далее--

```
adult=# SELECT * from ml_cat_predict ('boston.cbm', 'boston2');
row_num predict class
-----
0 24.99982028068538
1 20.664358727562394
2 33.67737911788664
3 34.28900239364565
4 34.61570849423551
5 27.968317495475695
6 21.68218578618033
```

Multi classification

adult=# SELECT * from astra3_predict;																	
row	alpha	delta	u	g	r	i	z	run_id	cam_col	field_id	spec_obj_id	redshift	plate	mjd	fiber_id	predict	class
1	16.9568897845004	3.64613008870454	23.33542	21.95143	20.48149	19.603	19.13094	7712	6	442	4.855016555329904e+18	0.5062369	4312	55511	495	0.98686	GALAXY
2	240.063240247767	6.13413059813973	17.86033	16.79228	16.43001	16.30923	16.25873	3894	1	243	2.4489280322708705e+18	0.0003448142	2175	54612	348	0.990419	STAR
3	30.887222067625	1.18870964120799	18.18911	16.89469	16.42161	16.24627	16.18549	7717	1	536	8.255357438959835e+18	4.085216e-06	7332	56683	943	0.997588	STAR
4	247.594400505002	10.8877797153666	24.99961	21.71203	21.47148	21.30532	21.29109	5323	1	134	4.577998722756271e+18	-0.0002914838	4066	55444	326	0.997667	STAR
5	18.8964507920807	-5.26133022886992	23.76648	21.79737	20.69543	20.23403	19.97464	7881	3	148	8.91047176642785e+18	-0.0001361561	7914	57331	363	0.996044	STAR
6	182.713733094955	51.3750050594777	22.44608	21.68444	20.24292	19.41423	19.08227	2830	1	411	7.516725588574623e+18	0.5026683	6676	56389	792	0.984373	GALAXY
7	150.089423193165	39.4670880748061	18.96441	17.82906	17.31429	16.99891	16.85583	3560	4	278	1.5267956411104236e+18	0.06366445	1356	53033	274	0.996164	GALAXY
8	189.510984338851	58.7411197772507	21.37376	20.80187	20.84925	21.13449	20.34689	2243	1	353	7.696817897528907e+18	0.7936153	6836	56443	604	0.957787	QSO
9	37.7138728560977	-0.525138228146508	20.77988	19.54618	19.16687	18.89438	18.64286	2700	2	117	1.7553283123029217e+18	0.1060118	1559	53271	183	0.993892	GALAXY
10	201.074980072746	28.7699058867715	25.05349	22.23362	20.8122	19.69488	19.28336	4649	3	120	7.306035245308205e+18	0.567082	6489	56329	257	0.993856	GALAXY
11	151.83091832672	19.8108624669417	24.04443	22.48608	20.59701	19.50985	19.00457	5183	5	142	6.622787444780849e+18	0.5475619	5882	56029	888	0.998885	GALAXY

adult=# SELECT * from ml_predict('astra3.cbm'		
'astra3');		
id	predict	class
0	0.9868595777513302	GALAXY
1	0.9904188657285139	STAR
2	0.9975875623929414	STAR
3	0.9976669380943318	STAR
4	0.9960439244920889	STAR
5	0.9843734017027631	GALAXY
6	0.9961635567874662	GALAXY
7	0.9577871819302538	QSO
8	0.9938922568658763	GALAXY
9	0.9938564131331261	GALAXY

Inner data model

```
adult=#
adult=# select name, j #> '{data_processing_options,cla
  name      |                class                | loss_func
-----+-----+-----+-----+-----+-----
astra      | ["GALAXY", "QSO", "STAR"] | "MultiClass"
titanic    | [0, 1]                      | "Logloss"
titanic    | [0, 1]                      | "Logloss"
boston     | []                          | "RMSE"
adult      | ["<=50K", ">50K"]          | "Logloss"
(5 rows)
```

PostgreSQL vs ClickHouse

```
adult=# select * from ml_predict('amazon.cbm' , 'amazon', '{
RESOURCE,
MGR_ID,
ROLE_ROLLUP_1,
ROLE_ROLLUP_2,
ROLE_DEPTNAME,
ROLE_TITLE,
ROLE_FAMILY_DESC,
ROLE_FAMILY,
ROLE_CODE}') )
LIMIT 10;
```

index	predict	class
0	5.075591747501174	1
1	4.677445251644691	1
2	3.4881006705946156	1
3	4.654735526605757	1
4	4.546219076437382	1
5	-0.7881046669169504	0
6	5.249330192285552	1
7	4.5361327711227215	1
8	4.542787758485275	1
9	3.9540183530568065	1

(10 rows)

```
catboostEvaluate('/tmp/amazon.cbm',
RESOURCE,
MGR_ID,
ROLE_ROLLUP_1,
ROLE_ROLLUP_2,
ROLE_DEPTNAME,
ROLE_TITLE,
ROLE_FAMILY_DESC,
ROLE_FAMILY,
ROLE_CODE) AS prediction,
^I ACTION AS target
FROM amazon_train LIMIT 10;


SELECT
catboostEvaluate('/tmp/amazon.cbm', RESOURCE, MGR
ACTION AS target
FROM amazon_train
LIMIT 10


Query id: c4975c2f-9380-4619-8b07-2dc9e3886470
```

prediction	target
5.075591747501174	1
4.677445251644691	1
3.4881006705946156	1
4.654735526605757	1
4.546219076437382	1
-0.7881046669169504	0
5.249330192285552	1
4.5361327711227215	1
4.542787758485275	1
3.9540183530568065	1


10 rows in set. Elapsed: 0.111 sec. Processed 8.19 th


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SQL 

1	25139	311198	91261	118026	122392	121143	173805	249618	121145
1	34924	28805	117961	118327	120299	124922	152038	118612	124924
1	80574	55643	118256	118257	117945	280788	280788	292795	119082
1	14354	59575	117916	118150	117920	118568	122142	19721	118570

93.228ms

4

```
1 SELECT pgml.transform(  
2   inputs => ARRAY[  
3     'I am Omar and I live in New York City.'  
4   ],  
5   task => 'token-classification'  
6 ) as ner;
```

ner
[[{"end":9,"entity":"I-PER","index":3,"score":0.9971067309379578,"start":5,"word":"Omar"},{"end":27,"entity":"I-LOC","index":8,"score":0.9993748068809508,"start":24,"word":"New"},{"end":32,"entity":"I-LOC","index":9,"score":0.9993545413017272,"start":28,"word":"York"},{"end":37,"entity":"I-LOC","index":10,"score":0.9994328618049622,"start":33,"word":"City"}]]

pg_ml vs PostgesML



Table "public.titanic"		
Column	Type	Collation
id	integer	
passenger_id	integer	
pclass	integer	
name	text	
sex	text	
age	double precision	
sibsp	integer	
parch	integer	
ticket	text	
fare	double precision	
cabin	text	
embarked	character(1)	
res	boolean	

convenient use of
categorical features

XGBoost

```
postgresml=# \d titanic
Table "public.titanic"
  Column      |      Type      | Collation |
-----+-----+-----+
 index        | bigint         |           |
 unnamed:0    | bigint         |           |
 pclass       | bigint         |           |
 name         | text           |           |
 sib_sp       | bigint         |           |
 parch        | bigint         |           |
 ticket       | text           |           |
 fare         | double precision |           |
 sex_male     | bigint         |           |
 nulls_1      | bigint         |           |
 nulls_2      | bigint         |           |
 cabin_mapped_1 | bigint         |           |
 cabin_mapped_2 | bigint         |           |
 cabin_mapped_3 | bigint         |           |
 cabin_mapped_4 | bigint         |           |
 cabin_mapped_5 | bigint         |           |
 cabin_mapped_6 | bigint         |           |
 cabin_mapped_7 | bigint         |           |
 cabin_mapped_8 | bigint         |           |
 embarked_q   | bigint         |           |
 embarked_s   | bigint         |           |
 survived     | bigint         |           |
Indexes:
    "ix_titanic_index" btree (index)
```