Summer Internship Presentation

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Assignment Details:

Develop a framework that can help benchmark multiple machine learning libraries

Background

Education



- Masters in Computer
 Science at State
 university of New York
 at Buffalo Graduating
 December 2016.
- Bachelor of Engineering in Computer Science.

Work Experience

- 4 years in Big Data Eco-space in companies -Impetus technologies, Center of Excellence at Cognizant.
- Worked on multiple Big data projects during the 4 years for clients like Capital One, American express, AT&T, Cisco, Lexus.
- 1 year in Java J2EE technologies.

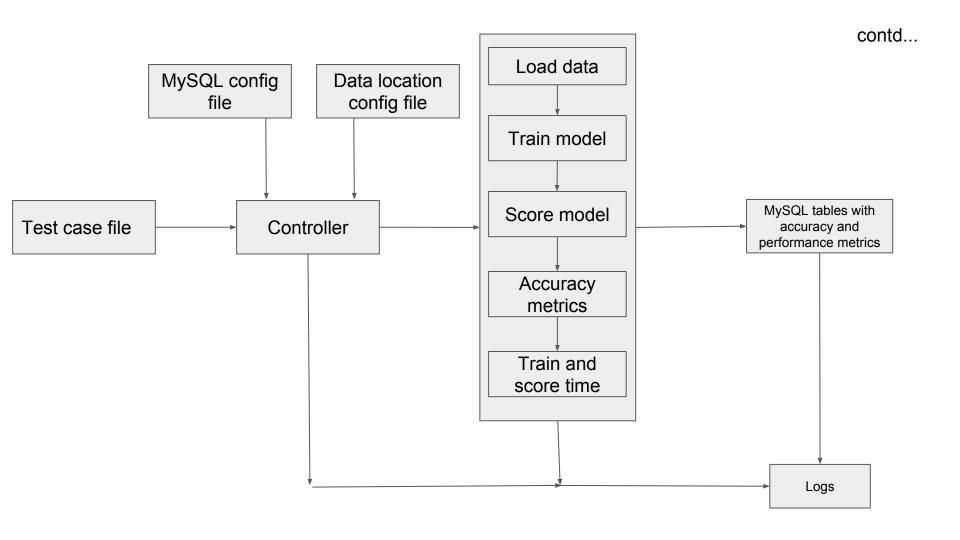
Project details

ML Frameworks used for benchmarking efforts:

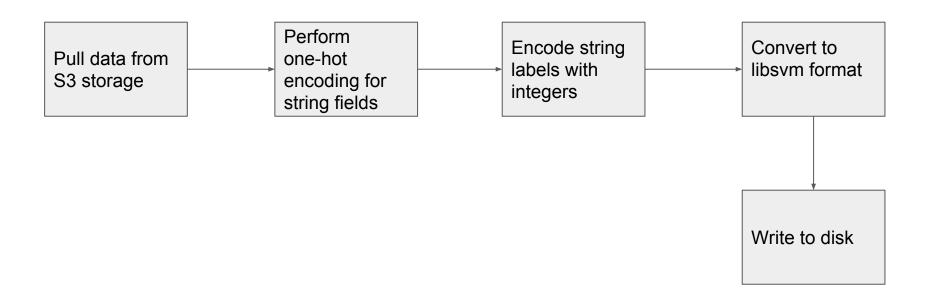
- Spark-ml
- Xgboost
- scikit

Benchmark workflow

- Read the test case file and get the Algorithm details and the hyper-parameters for the same.
- Connect to EC2 S3 storage and read the data files needed for the test case.
- Generate the feature matrix
- Run the algorithm
- Collect the accuracy metrics
- Collect the performance metrics training and scoring time
- Push the metrics to MySQL for each test case and each run
- Logs are written to a configurable directory



Preprocessing data for xgboost and scikit



Benchmark - MLlib

Algorithms supported:

- Linear regression
- Logistic regression Supports only binary classification.
- Random forest classifier
- Random forest regressor
- Gradient boosted machine classifier Supports only binary classification.
- Gradient boosted machine regressor
- Perceptron classifier
- KMeans
- Principal component analysis

Achieved through Java API's on Spark - 1.6.2

Benchmark - xgboost

Algorithms supported:

Gradient Boosting algorithm that supports regression and classification.

- A pre-processing step is needed which pulls data from S3, performs one-hot encoding and creates a libsvm file which is fed to xgboost
- Achieved through Java API's on xgboost 0.6

Benchmark - scikit learn

Algorithms supported:

- Linear regression
- Logistic regression
- Random forest classifier
- Random forest regressor
- Gradient boosted machine classifier
- Gradient boosted machine regressor
- Naive Bayes
- KMeans
- A pre-processing step is needed which pulls data from S3 and performs one-hot encoding
- Achieved through Python API's on scikit 0.17

Test cases file

```
test case id.algorithm,training data file id.test data file id.parameters.grid parameters.nfold,multiclass(T/F),header(Y/N),number of colmns if no header #100
1.glm.5.6.param regularizer:0.3:param elastic net:0.8:param number of iteration:10.param regularizer grid:0.1:0.5:param elastic net grid:0.1:1.nfold:0..N.90..
##2,glm,19,20,param_regularizer:0.3;param_elastic_net:0.8;param_number_of_iteration:10,param_regularizer_grid:0.1:0.5;param_elastic_net_grid:0.1:1,nfold:0,,Y,
3,qlm,21,22,param_regularizer:0.3;param_elastic_net:0.8;param_number_of_iteration:10,param_regularizer_grid:0.1:0.5;param_elastic_net_grid:0.1:1,nfold:0,,Y,,
4,qlm,23,24,param regularizer:0.3;param elastic net:0.8;param number of iteration:10,param regularizer grid:0.1:0.5;param elastic net grid:0.1:1,nfold:0,,Y,,
5.rfc.3.4.param number of trees:3:param max depth:10.param max depth grid:1:3:param number of trees grid:1:2.nfold:0.T.Y...
##6.rfc.5.6.param number of trees:3:param max depth:10.param max depth grid:1:3:param number of trees grid:1:2.nfold:0.N.N.90.
7,rfc,7,8,param_number_of_trees:3;param_max_depth:3;param_max_bins:110,param_max_depth_grid:1:3;param_number_of_trees_grid:1:2,nfold:0,T,Y,,
8,rfc,9,10,param number of trees:3;param max depth:10,param max depth grid:1:3;param number of trees grid:1:2,nfold:0,F,Y,,
##9,rfc,11,12,param_number_of_trees:3;param_max_depth:10,param_max_depth_grid:1:3;param_number_of_trees_grid:1:2,nfold:0,T,Y,55,
10, rfc, 13, 14, param_number_of_trees: 3; param_max_depth: 10, param_max_depth_grid: 1:3; param_number_of_trees_grid: 1:2, nfold: 0, T, Y, 785,
##15,rfc,1,2,param_number_of_trees:3;param_max_depth:10,param_max_depth_grid:1:3;param_number_of_trees_grid:1:2,nfold:0,T,Y,785,
##17,rfc,15,16,param number of trees:3;param max depth:10,param max depth grid:1:3;param number of trees grid:1:2,nfold:0,T,Y,785,
##18, rfc, 17, 18, param number of trees: 3; param max depth: 10, param max depth grid: 1:3; param number of trees grid: 1:2, nfold: 0, T, Y, 785,
19.rfr.5.6.param number of trees:3:param max depth:3.param max depth grid:1:3:param number of trees grid:1:2.nfold:0.N.N.90.
##20.rfr.19.20.param number of trees:3:param max depth:3.param max depth grid:1:3:param number of trees grid:1:2.nfold:0.N.Y..
21, rfr, 21, 22, param number of trees: 3; param max depth: 3, param max depth grid: 1:3; param number of trees grid: 1:2, nfold: 0, N, Y,,
22, rfr, 23, 24, param number of trees: 3; param max depth: 3, param max depth grid: 1:3; param number of trees grid: 1:2, nfold: 0, N, Y,,
##23.gbmr.5.6.param number of iteration:2:param max depth:2.param max depth grid:1:3.nfold:0.N.N.90.
##24,gbmr,19,20,param_number_of_iteration:2;param_max_depth:2,param_max_depth_grid:1:3,nfold:0,N,Y,,
25,qbmr,21,22,param_number_of_iteration:2;param_max_depth:2,param_max_depth_grid:1:3,nfold:0,N,Y,,
26,qbmr,23,24,param number of iteration:2;param max depth:2,param max depth grid:1:3,nfold:0,N,Y,,
27.gbmc.3.4.param number of iteration:2:param max depth:2.param max depth grid:1:3.nfold:0.F.Y..
28.dbmc.7.8.param number of iteration:2:param max depth:2:param max bins:110.param max depth grid:1:3.nfold:0.F.Y..
29, gbmc, 9, 10, param_number_of_iteration: 2; param_max_depth: 2, param_max_depth_grid: 1:3, nfold: 0, F, Y, , ,
##32,gbmc,15,16,param number of iteration:2;param max depth:2,param max depth grid:1:3,nfold:0,T,Y,,
##33.gbmc.17.18.param number of iteration:2:param max depth:2.param max depth grid:1:3.nfold:0.F.Y.,
34,nn,3,4,param_number_of_iteration:2;param_max_depth:2;param_layers_and_units:7#3#2,param_max_depth_grid:1:3,nfold:0,T,Y,,
35,nn,7,8,param_number_of_iteration:2;param_max_depth:2;param_layers_and_units:8#3#2,param_max_depth_grid:1:3,nfold:0,T,Y,,
##36,nn,9,10,param number of iteration:2;param max depth:2;param layers and units:10000#500#2,param max depth grid:1:3,nfold:0,T,Y,,
37,nn.11,12,param number of iteration:2;param max depth:2;param layers and units:54#27#7,param max depth grid:1:3,nfold:0,T,Y,55,
38, nn, 14, 14, param_number_of_iteration: 2; param_max_depth: 2; param_layers_and_units: 784#350#10, param_max_depth_grid: 1:3, nfold: 0, T, Y, 785,
##39,nn,15,16,param_number_of_iteration:2;param_max_depth:2;param_layers_and_units:7#3#2,param_max_depth_grid:1:3,nfold:0,T,Y,,,
##40,nn,17,18,param number of iteration:2;param max_depth:2;param layers_and_units:7#3#2,param_max_depth_grid:1:3,nfold:0,T,Y,,
41, qlmlogistic, 3,4, param regularizer: 0.3; param elastic net: 0.8; param number of iteration: 10, param regularizer grid: 0.1: 0.5; param elastic net grid: 0.1:1, nfold: 0.1:0, param regularizer grid: 0.1:0, param regularizer: 0.3; param elastic net: 0.8; param number of iteration: 10, param regularizer grid: 0.1:0, param regularizer: 0.3; param elastic net: 0.8; param number of iteration: 10, param regularizer: 0.3; param elastic net: 0.8; param number of iteration: 10, param regularizer: 0.1:0, param regular
42.olmlogistic.7.8.param regularizer:0.3:param elastic net:0.8:param number of iteration:10.param regularizer grid:0.1:0.5:param elastic net grid:0.1:1.nfold:0
##43.glmlogistic.10.10.param regularizer:0.3:param elastic net:0.8:param number of iteration:10.param regularizer grid:0.1:0.5:param elastic net grid:0.1:1.nfg
##44,glmlogistic,17,18,param regularizer:0.3;param elastic net:0.8;param number of iteration:10,param regularizer_grid:0.1:0.5;param elastic net grid:0.1:1.1,nfc
##45, kmeans, 9, 10, param_number_of_iteration: 2; param_number_of_clusters: 5,,
#TEST CASES WITH CROSS VALIDATION
#GBT supports only binary classification
```

Data file details

```
id,data path,feature matrix columns,output column name,feature matrix column if string,label column name
1,s3n://h2o-public-test-data/smalldata/testng/iris train1.csv,Sepal.Length;Sepal.Width;Petal.Length,features,.Petal.Width
2.s3n://h2o-public-test-data/smalldata/testng/iris validation1.csv
3,s3n://h2o-public-test-data/smalldata/testng/prostate_train.csv,AGE;RACE;DPROS;DCAPS;PSA;VOL;GLEASON,features,,CAPSULE
4.s3n://h2o-public-test-data/smalldata/testng/prostate test.csv
5,s3n://h2o-public-test-data/bigdata/laptop/testng/milsongs-train.csv.qz,,test,,,
6,s3n://h2o-public-test-data/biqdata/laptop/testng/milsongs-test.csv.qz
7,s3n://h2o-public-test-data/smalldata/testng/airlines_train.csv,Distance,features,fYear;fMonth;fDayofWooth;fDayofWeek;UniqueCarrier;Origin;Dest,IsDepDelayed
8,s3n://h2o-public-test-data/smalldata/testng/airlines_test.csv
9,s3n://h2o-public-test-data/smalldata/testng/arcene train.csv,X1;C2;C3;C4;C5;C6;C7;C8;C9;C10;C11;C12;C13;C14;C15;C16;C17;C18;C19;C20;C21;C22;C23;C24;C25;C26;C27;C28;C29;C30;C3
11,s3n://h2o-public-test-data/smalldata/testng/covtype_small_dense_multiclass_unbalanced_train.csv
12.s3n://h2o-public-test-data/smalldata/testng/covtype small dense multiclass unbalanced test.csv
13.s3n://h2o-public-test-data/biqdata/laptop/testnq/mnist train.csv
14,s3n://h2o-public-test-data/bigdata/laptop/testng/mnist_test.csv
15,s3n://h2o-public-test-data/bigdata/laptop/testng/cup98_train.csv
16,s3n://h2o-public-test-data/bigdata/laptop/testng/cup98_test.csv
17,s3n://h2o-public-test-data/bigdata/laptop/testng/higgs_train_imbalance_100k.csv
18.s3n://h2o-public-test-data/bigdata/laptop/testng/higgs_test_imbalance_100k.csv
19,s3n://h2o-public-test-data/smalldata/testng/cars train.csv.economy (mpg);cylinders;power (hp);displacement (cc);weight (lb);year,features,name,0-60 mph (s)
20.s3n://h2o-public-test-data/smalldata/testng/cars test.csv
21.s3n://h2o-public-test-data/smalldata/testng/housing train.csv.C1:C2:C3:C4:C5:C6:C7:C8:C9:C10:C11:C12:C13.features..C14
22.s3n://h2o-public-test-data/smalldata/testng/housing test.csv
23,s3n://h2o-public-test-data/smalldata/testng/computer_train.csv,C3;C4;C5;C6;C7;C8,features,,C10
24,s3n://h2o-public-test-data/smalldata/testng/computer_test.csv
25.s3n://h2o-public-test-data/smalldata/testng/iris.csv
26,s3n://h2o-public-test-data/smalldata/testng/prostate.csv
27,s3n://h2o-public-test-data/bigdata/laptop/testng/milsongs.csv.gz
28,s3n://h2o-public-test-data/smalldata/testng/airlines.csv
29,s3n://h2o-public-test-data/smalldata/testng/arcene.csv
30.s3n://h2o-public-test-data/smalldata/testng/covtype small dense multiclass unbalanced.csv
31,s3n://h2o-public-test-data/bigdata/laptop/testng/mnist.csv
32,s3n://h2o-public-test-data/bigdata/laptop/testng/cup98.csv
33.s3n://h2o-public-test-data/bigdata/laptop/testng/higgs imbalance 100k.csv
34,s3n://h2o-public-test-data/smalldata/testng/cars.csv
35,s3n://h2o-public-test-data/smalldata/testng/housing.csv
36,s3n://h2o-public-test-data/smalldata/testng/computer.csv
```

MySQL config

```
driver = com.mysql.jdbc.Driver
db = h2o
host = 172.16.2.178
user = root
password = 0xdata
table = SparkStats
port = 3306
```

MySQL metrics table

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- 1,	glm		NULL		NULL		NULL		NULL			NULL		NULL		NULL	Regression	1
- 4				0.43884691		19.04954		588		31			s3n://n2o-pul			ng/computer_train.csv	D	
	glm	NULL	NULL	NULL	NULL	NULL	NULL	I 3373	NULL	- 1		NULL	-2 / /h2	NULL	-data/smalldata/test	NULL	Regression	
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		7.33010390					3158888475									/testng/milsongs-train		1
	glm		NULL		NULL	10.00144	NULL		NULL	34		NULL	5511.//1120-pui	NULL I		NULL	Regression	1
		NULL	NOLL	NULL	NOLL	NULL	INOLL	9127		57 1			s3n · //h2n-nul			ng/airlines_train.csv	Regression	
	rfc		0.5665018		0.61058105		1 0.6105810	572867614					310572867614			NULL	Classification	1
		NULL	0.5005010	NULL	0.01050105	NULL	0.0103010	63442							-data/smalldata/test		c tubbilized tion	1
	rfc		NULL		NULL	HOLL	NULL		NULL	-0		NULL	33111//1120 pui			0.9684659090909091	Classification	1
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ī	rfc		0.8999811	145775389	0.90005		0.90005	1			16891			NULL I		I NULL	Classification	T
19		7.88341941		0.10409521		10.59113	31554352721	19153								testng/milsongs-train		4
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21	1	45.6748657		0.16188355	038473445	56.15438	3219928148	1 265		28	147094090	1104	s3n://h2o-pul	olic-test	-data/smalldata/test	ng/housing train.csv		
1	rfr	.	NULL	1	NULL		NULL		NULL		1.1	NULL		NULL I		NULL	Regression	I
22	2	13.4475353	90062225	0.56217245	23771108	16.82656	7953132106	219	1 2	27	147094091	7420	s3n://h2o-pul	olic-test	-data/smalldata/test	ng/computer_train.csv		
- 1	rfr	. 1	NULL	1	NULL		NULL		NULL		1.1	NULL		NULL	NULL	NULL	Regression	1
25	5	35.7103498	5861266	0.47059963	360347745	44.62966	120882825	15299	1 2	28	147094092	1737	s3n://h2o-pul	olic-test	-data/smalldata/test	ng/housing_train.csv		
- 1	gbm	ır	NULL	1	NULL		NULL		NULL			NULL		NULL	NULL	NULL	Regression	1
26	5	7.13248320	2426509	0.81747838	45356943	10.86428	35104581114	16513	1 2	27	147094096	0023	s3n://h2o-pul	olic-test	-data/smalldata/test	ng/computer_train.csv		
1	gbm	ir	NULL	1	NULL		NULL	1	NULL		1	NULL		NULL	NULL	NULL	Regression	1
27	7	NULL		NULL		NULL		32174	1 2	23 1	147094098	4230	s3n://h2o-pul	olic-test	-data/smalldata/test	ng/prostate_train.csv		
	gbm		NULL		NULL		NULL		NULL			NULL		NULL	0.8439449296025664	0.8561435414990957	Classification	1
28	3	NULL		NULL		NULL		8420		25 1			s3n://h2o-pul		-data/smalldata/test			
	gbm		NULL		NULL		NULL	1	NULL			NULL			0.6669995513119844		Classification	1
		NULL		NULL		NULL		55948		32			s3n://h2o-pul		-data/smalldata/test			
	gbm		NULL		NULL		NULL		NULL			NULL			1.0	1.0	Classification	1
35		NULL		NULL		NULL		31386							-data/smalldata/test			
-	nn		0.3906508	3272825593	0.54715204		0.5471520	412759511					52041275951			NULL	Classification	
		NULL		NULL		NULL		7973								ng/covtype_small_dense		
	nn		0.7578336	869011121	0.83351648		0.8335164	835164836					164835164836		NULL	NULL	Classification	
		NULL		NULL		NULL		89785								testng/mnist_test.csv		
	nn		0.5454015	5569671474			0.5912		0.6581576				000000000001			NULL	Classification	1
		NULL	0 70000	NULL		NULL	1	359		34			s3n://h2o-pul			ng/prostate_train.csv	61	
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	nn		0.4420292	1882345	0.59344262	229508196	0.5934426	229508196	0.3521/4	140/3	4/480/	0.5934	126229508196	NULL	NULL	NULL	Classification	1

contd..

mysgl> select * from XgboostStats: | run | trainrmse | testrmse | trainingae | testmae | trainlogloss | testlogloss | trainrerror | testerror | trainauc | testauc | trainingtime | testtime | timestamp | dataset 1 | 1 | 0.168564 | 0.170865 | 0.142931 | 0.144147 | 0.162225 0.164609 0.007525 0 | 1470946864967 | /Users/nikhilshekha 0.007318 | 0.997955 | 0.997318 | r/h2o/xgboost/demo/data/agaricus.txt.train l abm gbm 1 | 2 | 0.075735 | 0.080230 | 0.053599 | 0.054797 | 0.059073 0.061924 0.002098 0.002918 0.999014 | 0.998326 | 0 | 1470946864967 | /Users/nikhilshekha r/h2o/xgboost/demo/data/agaricus.txt.train | gbm | qbm 1 | 3 | 0.041020 | 0.047022 | 0.021570 | 0.022513 | 0.024808 0.027473 0.001024 0.001843 84 | 0 | 1470946864967 | /Users/nikhilshekha | 0.999065 | 0.998697 | r/h2o/xgboost/demo/data/agaricus.txt.train l abm gbm 1 | 4 | 0.027302 | 0.033732 | 0.009101 | 0.009886 | 0.011760 0.014158 0.000460 1 0.001382 | 0.999171 | 0.998816 | 84 | 0 | 1470946864967 | /Users/nikhilshekha r/h2o/xgboost/demo/data/agaricus.txt.train gbm qbm 1 | 5 | 0.023020 | 0.027296 | 0.004178 | 0.004737 | 0.006097 0.007927 0.000460 0.000921 | 0.999806 | 0.999552 | 0 | 1470946864967 | /Users/nikhilshekha r/h2o/xgboost/demo/data/agaricus.txt.train gbm 2 | 1 | 0.416912 | 0.421947 | 0.360991 | 0.366655 | 0.522843 0.535534 0 | 1470946865906 | /Users/nikhilshekha 0.263100 1 0.292763 0.818024 | 0.803852 | 15 I r/h2o_working_directory/test/s3data_new/prostate_train.lib | qbm | gbm 2 | 2 | 0.404101 | 0.412338 | 0.330628 | 0.336606 | 0.495632 0.521224 0.232533 0.233553 | 0.834429 | 0.818020 | 15 I 0 | 1470946865906 | /Users/nikhilshekha r/h2o working directory/test/s3data new/prostate train.lib | gbm I abm 0 | 1470946865906 | /Users/nikhilshekha 2 | 3 | 0.402385 | 0.413338 | 0.326205 | 0.333154 | 0.491851 0.524826 0.222707 0.233553 | 0.833742 | 0.811651 | 15 I r/h2o_working_directory/test/s3data_new/prostate_train.lib | gbm gbm 2 | 4 | 0.402228 | 0.413681 | 0.325030 | 0.332208 | 0.491452 0.525749 0.225982 0.240132 | 0.834279 | 0.811779 | 15 | 0 | 1470946865906 | /Users/nikhilshekha r/h2o working directory/test/s3data new/prostate train.lib | gbm I abm 2 | 5 | 0.402218 | 0.413716 | 0.324860 | 0.332081 | 0.491422 0.525792 0.223799 0.240132 | 0.834097 | 0.810320 | 15 I 0 | 1470946865906 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/prostate_train.lib | gbm gbm 4 | 1 | 1.155843 | 76.990936 | 0.458816 | 57.607040 | -6051.563477 | -6010.870117 | -164.260040 | -163.155487 | NULL NULL 8 | 0 | 1470946866606 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/cars_train.lib | qlm | qlm 4 | 2 | 1.150747 | 76.994400 | 0.456809 | 57.607822 | -6051.563477 | -6010.870117 | -164.260040 | -163.155487 | NULL I NULL 0 | 1470946866606 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/cars_train.lib I alm I alm 8 | 4 | 3 | 1.145684 | 76.997871 | 0.454821 | 57.608604 | -6051.563477 | -6010.870117 | -164.260040 | -163.155487 | NULL I NULL 0 | 1470946866606 | /Users/nikhilshekha r/h2o working directory/test/s3data new/cars train.lib qlm | qlm 4 | 4 | 1.140651 | 77.001343 | 0.452839 | 57.609375 | -6051.563477 | -6010.870117 | -164.260040 | -163.155487 | NULL I NULL 8 | 0 | 1470946866606 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/cars_train.lib alm I alm 4 | 5 | 1.135651 | 77.004807 | 0.450869 | 57.610149 | -6051.563477 | -6010.870117 | -164.260040 | -163.155487 | NULL 8 1 NULL 0 | 1470946866606 | /Users/nikhilshekha r/h2o working directory/test/s3data new/cars train.lib glm | qlm 5 | 1 | 7.475157 | 7.574760 | 4.887924 | 4.941753 | -789.414856 | -791.454346 | -21.432821 | -21.482767 I NULL 10 | 0 | 1470946869118 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/housing_train.lib | glm I alm 5 | 2 | 6.901058 | 7.000740 | 4.499975 | 4.571213 | -789.762329 | -791.454346 | -21.440079 | -21.482767 NULL NULL 10 | 0 | 1470946869118 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/housing_train.lib | qlm | qlm 5 | 3 | 6.506752 | 6.598309 | 4.275152 | 4.344232 | -792.080139 | -791.454346 | -21.499920 | -21.482767 NULL NULL 10 | 0 | 1470946869118 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/housing_train.lib | glm I alm 5 | 4 | 6.225984 | 6.312995 | 4.145546 | 4.211830 | -793.167786 | -793.388489 | -21.529274 | -21.544903 I NULL 10 | 0 | 1470946869118 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/housing_train.lib | qlm | qlm 5 | 5 | 6.021369 | 6.107774 | 4.071876 | 4.132611 | -793.411804 | -795.514038 | -21.529274 | -21.592960 NULL NULL 10 | 0 | 1470946869118 | /Users/nikhilshekha r/h2o working directory/test/s3data new/housing train.lib | glm I alm 6 | 1 | NULL I NULL I NULL I NULL NULL I NULL I NULL NULL I NULL I NULL 76934 | 0 | 1470946870537 | /Users/nikhilshekha r/h2o_working_directory/test/s3data_new/mnist_train.lib | classifier | classifier | 6 | 2 | NULL NULL NULL NULL NULL NULL NULL NULL NULL 76934 0 | 1470946870537 | /Users/nikhilshekha

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Learning outcomes

- Deep exposure to h2o, Spark ML and xgboost machine learning libraries.
- Stumbled upon multiple unresolved JIRA's for spark and had to look for alternative solution to solve the problem at hand.
- Exposure to multiple datasets stored in different formats.
- Dealing with missing labels, fields in Spark-ml, xgboost, scikit.
- Reading data from s3 buckets programmatically using Java and Python.

Work in progress

- Run Spark in multi-node environment (Our own machines to start with, but EC2 if we ever decide to publish this framework)
- Run xgboost in a distributed environment with spark (Code is available in "ml-benchmark" repo, but needs testing)
- Tune the hyper-parameters for optimal performance of algorithms
- Add more test cases to each of the libraries (Utilize more "Kaggle like" datasets. Right now we have whatever is available in S3 for H2O)
- Run algorithms on large datasets (Besides the airlines dataset...)
- Integrating h2o accuracy suite First, need to publish the test classes to maven central repository.
 - Work around is to paste in necessary classes into the "ml-benchmark" repo for H2O
- Benchmark.ai ? -> A combination of "ml-benchmark" and "db-benchmark"
- Other frameworks to benchmark?

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Thank You