# **Project Report—Analyze Models**

#### **Get news from New York Times**

python news\_api.py

#### Compare binary classifiers when number of Features is 200

bin/spark-submit tfidf\_classify\_spark.py

**DATA:** restuarant reviews {Positive; Negative}

When number of features is 200:

Compare the confusion matrix and test error rate of models

{NaiveBayes; SVMWithSGD; LogisticRegressionWithBFGS; OnlinePerceptron(Itr=1); AveragePerceptron(Itr=1); OnlinePerceptron(Itr=1)+AveragePerceptron(Itr=1); AveragePerceptron(Itr=5); AveragePerceptron(Itr=10)}

```
Yilans-MacBook-Air:spark-2.0.1-bin-hadoop2.7 Jillian$ bin/spark-submit ~/Documents/Python/large_data_pj/tfidf_classify_spark.py
Test error rate of NaiveBayes Model: 0.236236236236
Test error rate of SVMWithSGD Model: 0.253253253253
Test error rate of LogisticRegressionWithLBFGS Model: 0.208208208208
Test error rate of OnlinePerceptron: 0.29029029029
Test error rate of AveragePerceptron(iterations=1): 0.232232232232
Test error rate of AveragePerceptron(iterations=1) after single-pass OnlinePerceptron: 0.205205205205
Test error rate of AveragePerceptron(iterations=5):: 0.203203203203
Test error rate of AveragePerceptron(iterations=10): 0.212212212212
Confusion Matrix of models:
                                                |falseNegative|falsePositive|trueNegative|truePositive|
Model
                            |errorRate
|NaiveBayes
                            [0.23623623623623624]83
                                                              234
                                                                                         656
                            [0.25325325325325326]19
                                                                             90
|LogisticRegressionWithLBFGS|0.2082082082082082 |82
                                                               126
                                                                             198
                                                                                          593
                            [0.2902902902902903 [207
                                                                             241
                                                                                          468
|OnlinePerceptron
                            0.23223223223223224|80
                                                                             172
                                                                                          595
AveragePerceptronItr1
                                                               152
                            0.20520520520520522|78
                                                               127
AveragePerceptronOPItr1
                                                                             197
                                                                                          597
                            0.2032032032032032
                                                                             197
                                                                                          599
AveragePerceptronItr5
                                                                             191
                                                                                          596
AveragePerceptronItr10
                            [0.2122122122122122 [79
                                                              133
```

#### Compare binary classifiers when number of Features is 2000

bin/spark-submit tfidf\_classify\_spark.py

**DATA:** restuarant reviews {Positive; Negative}

When number of features is 2000:

Compare the confusion matrix and test error rate of models

{NaiveBayes; SVMWithSGD; LogisticRegressionWithBFGS; OnlinePerceptron(Itr=1); AveragePerceptron(Itr=1); OnlinePerceptron(Itr=1)+AveragePerceptron(Itr=1); AveragePerceptron(Itr=5); AveragePerceptron(Itr=10)}

```
Yilans-MacBook-Air:spark-2.0.1-bin-hadoop2.7 Jillian$ bin/spark-submit ~/Documents/Python/large_data_pj/tfidf_classify_spark.py
Test error rate of NaiveBayes Model: 0.159159159159
Test error rate of SVMWithSGD Model: 0.223223223223
Test error rate of LogisticRegressionWithLBFGS Model:
Test error rate of OnlinePerceptron: 0.14014014014
Test error rate of AveragePerceptron(iterations=1): 0.17017017017
Test error rate of AveragePerceptron(iterations=1) after single-pass OnlinePerceptron: 0.106106106106
Test error rate of AveragePerceptron(iterations=5):: 0.115115115115
Fest error rate of AveragePerceptron(iterations=10): 0.103103103103
Confusion Matrix of models:
Model
                                                |falseNegative|falsePositive|trueNegative|truePositive|
                            0.15915915915915915|63
NaiveBayes
                                                                                          612
                                                              205
SVMWithSGD
                            [0.22322322322322322]18
                                                                             119
                                                                                          657
                                                                             324
                                                                                          675
LogisticRegressionWithLBFGS[0.0]
                                                               0
                            0.14014014014014014015 47
                                                               93
                                                                             231
OnlinePerceptron
AveragePerceptronItr1
                            [0.17017017017017017]58
                                                               112
                                                                             212
                                                                                          617
AveragePerceptronOPItr1
                            [0.1061061061061061 [47
                                                                                          628
                                                               59
                                                                             265
                                                               77
                            [0.11511511511511512]38
                                                                             247
AveragePerceptronItr5
                            0.1031031031031031 | 36
                                                                             257
                                                                                          .
| 639
AveragePerceptronItr10
```

## **MulticlassPerceptron Training error rate and Confusion Matrices:**

bin/spark-submit category\_news\_train\_perceptron.py

Data: News from NEW YORK TIMES in 2016 year, {number of classes=19; number of features=2000}

Lifting perceptron binary class models to MulticlassPerceptron using One-vs-Rest

## **Training error rate**

	r:spark-2.0.1-bin-hac ain_perceptron.py	doop2.7 Jillians	bin/spark-su	bmit ~/Docume	ents/Python/lar	ge_da
category	errorRate	falseNegative	falsePositive	trueNegative	truePositive	
Art & Design	0.10457633982810602	2	2322	19200	699	
World	0.09116680916167934	180	1846	15951	4246	
Sports	0.04949826756063538	62	1038	16968	4155	
Fashion & Style	0.1538946136885209	2	3418	18055	748	
Books	0.029788957386491472	1	661	20360	1201	
Music	0.06929757458488953	1	1539	19601	1082	
Television	0.04967826126085587	1	1103	20588	531	
Movies	0.03644872429464969	2	808	20400	1013	
Technology	0.11515096971605994	0	2559	18871	793	
Science	0.13990010349637763	0	3109	18505	609	
Food	0.04396346127885524	0	977	20727	519	
Real Estate	0.0559780407685731	0	1244	20626	353	
Theater	0.06754263600773973	0	1501	19998	724	
Health	0.11141610043648473	2	2474	18819	928	
Travel	0.0949916752913648	0	2111	19773	339	
Education	0.07397741079062232	0	1644	19910	669	
Your Money	0.07555235566755164	0	1679	20388	156	
Politics	0.05858794942177024	19	1283	17929	2992	
Economy	0.07109751158709446	0	1580	20449	194	
+		·			++	

#### **Compare Multiclass Models Test error rate and Confusion Matrices:**

bin/spark-submit compare\_Multiclass\_Models.py

Data: News from NEW YORK TIMES in 2016 year, {number of classes=19; number of features=2000}

```
Yilans-MacBook-Air:spark-2.0.1-bin-hadoop2.7 Jillian$ bin/spark-submit ~/Documents/Python/large_data_pj
/compare_Multiclass_Models.py
training data set count: 17765
test data set count: 4458
NaiveBayes Test set accuracy = 0.814266487214
DecisionTree Test set accuracy = 0.411395244504
RandomForest model: RandomForestClassificationModel (uid=rfc_56a2c3d1ed00) with 10 trees
RandomForest Test set accuracy = 0.449977568416
One-vs-Rest (LogisticRegression) Test set accuracy = 0.758187528039
MulticlassPerceptron accuracy:
                             errorRate|falseNegative|falsePositive|trueNegative|truePositive|
        category|
   Art & Design| 0.09510991475998205|
                                                                412|
                                                                             3902|
                                                                                            132|
                                                  12|
          World 0.11799013010318528
                                                  103|
                                                                423
                                                                             31411
                                                                                            791
          Sports| 0.07514580529385374|
                                                  66|
                                                                2691
                                                                             33311
                                                                                            792
 Fashion & Style| 0.16464782413638404|
                                                                698 j
                                                                                            123
                                                   361
                                                                             3601
           Books | 0.036787797218483624
                                                   18
                                                                146
                                                                             4061
                                                                                            233
           Music| 0.09914759982054733|
                                                   23 |
                                                                4191
                                                                             38281
                                                                                            188
      Television[0.048900852400179454]
                                                    91
                                                                 2091
                                                                             41501
                                                                                            90
          Movies | 0.03297442799461642|
                                                   161
                                                                131
                                                                             41241
                                                                                            187
     Technology | 0.13593539703903096 |
                                                   191
                                                                 5871
                                                                             3713|
                                                                                            139
         Science 0.15029161058770749
                                                   21 |
                                                                6491
                                                                             36791
                                                                                            109
            Food | 0.05361148497083894|
                                                   11|
                                                                 228
                                                                             4117
                                                                                            102
    Real Estate | 0.11238223418573351 |
                                                    5 |
                                                                496
                                                                             3897
                                                                                            60
         Theater| 0.04441453566621804|
                                                                                            134
                                                    9
                                                                189
                                                                             4126
          Health 0.13885150291610587
                                                                602
                                                                             3662
                                                   17
          Travel| 0.09062359802602064|
                                                    6
                                                                398
                                                                             3995
                                                                                            59
       Education | 0.06370569762225213|
                                                                             4051
                                                    9|
                                                                 275
                                                                                            123
      Your Money | 0.06729475100942127
                                                                 297
                                                    3 [
                                                                             4133
        Politics 0.07447285778375953
                                                                             3600
                                                                                            526
                                                                 290
         Economy | 0.08613728129205922|
                                                                             4035
                                                                                             39
                                                                 3801
```

### Compare 4 MulticlassPerceptron Models with different initial weight

bin/spark-submit category\_tweets\_perceptron.py --master local[10]

Training data is Feedback of correct categories of tweets (200)

Predict categories of test tweets using models after online trained every 50 tweets

- OnlinePerceptron model with weight initialized as zeros
- AveragePerceptron model with weight initialized as zeros
- AveragePerceptron model with weight initialized as trained model by news(but ignore weights history of 22223 news)
- AveragePerceptron model with weight initialized as trained model by news(and store weights history of 22223 news)

```
print "Training 4 MulticlassPerceptron models using",dataset.count(),"training data..........\n"
# train an online perceptron model use part of training data
initialModelOP.train(traindata,trainlabels,method="Online")
# train an Average perceptron model(Iteration=10) use part of training data
#initialModelAP.train(traindata,trainlabels,method="Average",source="News")
# train an Average perceptron model(Iteration=10) use part of training data
loadModelOP.train(traindata,trainlabels,method="Average",source="Feedback")
# train an Average perceptron model(Iteration=10) use part of training data
loadModelAP.train(traindata,trainlabels,method="Average",source="Feedback")

# train an Average perceptron model(Iteration=10) use part of training data
loadModelAP.train(traindata,trainlabels,method="Average",source="Feedback")

print "Predicting labels of test Tweets after trained ",dataset.count()," Category Feedback Tweets:\n"
```

[^[[AYilans-MacBo	ook-Air:spark-2.0.1-bin-hadoop2.7 Jillian\$ bin/spark-submit ~/Documents/Python/large_data_pj/category_tweets_perceptron.py
category	text
Others  Others  Food  Others  Your Money  Others	really hate folks w. pissy ass attitudes for nun shit no where near cute   " even ignoring @RuPaul in her quest to become a mindless mogul." Never! https://t.co/qr4mecWmSv   Bout to eat at CB
Others  Politics  Others  Travel  Others  Fashion & Style	"Where dreams are made of" is not poetic license, it's just bad grammar.   Soldiers dying in war with naksals, party ministers and people busy celebrating delhi MCD election, not good !   Bad little bitch, wanna know how her lips taste   We have these in downtown Omaha, Nebraska ⊕https://t.co/hvUg9PoqGM     It's weird that you can remember dreams, that make no sense. But for some reason it will stick with you.   @Phil_Lewis_ Next up, underwear with artificial skid marks.   @clairemacckyd Are you referring to the dinked middle? Do you have 7" "spider" adapter? Or did you mean something e… https://t.co/2bCHaynfsF
1	If u put actual clothes on & do your makeup everyday before class, what r u doing?!?!?   @Haylie_Ewald3 so jealous
only showing top	20 rows

AveragePerceptron	NewsPerceptronIgnoreWeight	NewsPerceptronStoreWeight	OnlinePerceptror
Others :	Sports Fashion & Style Health Travel	Sports Fashion & Style Health Travel	Others
Others	Fashion & Style Movies Health Travel	Fashion & Style Movies Health Travel	Others
Others	Sports Fashion & Style Technology Science Health Education	Sports Fashion & Style Technology Science Health Education	Others
Others	Art & Design Sports Health Education	Art & Design Sports Health Education	Others
Others	Music Movies Real Estate Travel	Music Movies Real Estate Travel	Others
Others	World Health	World Health	Others
Others	Art & Design Fashion & Style Food Health Travel Your Money Economy	Art & Design Fashion & Style Food Health Travel Your Money Economy	Others
Others	World Fashion & Style Travel	World Fashion & Style Travel	Others
Others	Sports Fashion & Style Music	Sports Fashion & Style Music	Others
Others	World Health	World Health	Others
Others	Sports Science Health	Sports Science Health	Others
Others	Fashion & Style Music Real Estate	Fashion & Style Music Real Estate	Others
Others :	Science Education	Science Education	Others
Others [	Technology Health Education Your Money	Technology Health Education Your Money	Others
Others	Fashion & Style Education	Fashion & Style Education	Others
Others	Fashion & Style Music Science Health	Fashion & Style Music Science Health	Others
Others	World	World	Others
Others	Others	Others	Others
Others	World Sports Fashion & Style Health Your Money Economy	World Sports Fashion & Style Health Your Money Economy	Others
Others i	World Sports Health	World Sports Health	Others

Training 4 MulticlassPerceptron models using 50 training data.....

Training 4 MulticlassPerceptron models using 50 training data.....

Predicting labels of test Tweets after trained 50 Category Feedback Tweets:

	+	.+	.+
AveragePercep	tron NewsPerceptronIgnoreWeight	NewsPerceptronStoreWeight	OnlinePerceptron
Others	Health Travel	Fashion & Style Health Travel	Others
Others		Fashion & Style Movies Health Travel	Others
Others	Science Food	Sports Fashion & Style Technology Science	Others
Others	Sports	Sports Health Education	Others
Others	Music Movies Travel	Music Movies Travel	Others
Others	Others	World	Others
Others	Others	Art & Design Fashion & Style Food Health Econom	v Others
Others	World Fashion & Style	World Fashion & Style Travel	Others
Others	Fashion & Style	Sports Fashion & Style Music	Others
Others	World	World Health	Others
Others	Sports	Sports Science Health	Others
Others	Music	Music	Others
Others	Science Education	Science Education	Others
Others	Fashion & Style	Technology Health Education	Others
Others	Others	Fashion & Style	Others
Others	Music	Music Health	Others
Others	World	World	Others
Others	Others	Others	Others
Others	World Sports Fashion & Style Econom	y Sports Fashion & Style Health Your Money Econom	
Others	World	World Sports Health	Others

only showing top 20 rows

Training 4 MulticlassPerceptron models using 48 training data.....

Training 4 MulticlassPerceptron models using 48 training data.....

Predicting labels of test Tweets after trained 48 Category Feedback Tweets:

AveragePerce	ptron NewsPerceptronIgnoreWeight	NewsPerceptronStoreWeight	OnlinePerceptron
Others	Travel	Årt & Design Sports Fashion & Style Technology Health Travel	Others
Others	Fashion & Style Movies Travel	World Fashion & Style Movies Health Travel	Others
Others	Science Food	Sports Fashion & Style Technology Science Health Education	Others
Others	Others	Art & Design Sports Fashion & Style Movies Health Education	Others
Others	Others	Music Movies Travel	Others
Others	Others	World Science Health	Others
Others	Others	Art & Design Fashion & Style Science Food Health Travel Economy	Others
Others	World Fashion & Style	World Fashion & Style Travel	Others
Others	Fashion & Style	Sports Fashion & Style Music Technology	Others
Others	World	World Health	Others
Others	Sports	Sports Technology Science Health	Others
Others	Music	Sports Fashion & Style Music Science Real Estate Health	Others
Others	Education	Technology Science Education	Others
Others	Fashion & Style	Technology Health Education	Others
Others	Others	Fashion & Style Technology Health Education	Others
Others	Music	Sports Fashion & Style Music Technology Health	Others
Others	World	World Health Education	Others
Others	Others	Sports	Others
Others	World Fashion & Style Economy	Sports Fashion & Style Health Your Money Economy	Others
Others	World	World Sports Health Education	Others

only showing top 20 rows

Training 4 MulticlassPerceptron models using 48 training data.....

Fraining 4 MulticlassPerceptron models using 48 training data..... Predicting labels of test Tweets after trained 48 Category Feedback Tweets: |AveragePerceptron|NewsPerceptronIgnoreWeight |NewsPerceptronStoreWeight |OnlinePerceptron| Others |Art & Design Sports Fashion & Style Technology Science Real Estate Health Travel Education|Others Others |Fashion & Style Movies Food Travel|World Fashion & Style Movies Technology Health Travel Education Others (Science Food |Art & Design Sports Fashion & Style Technology Science Health Education Others Others |Sports |Art & Design Sports Fashion & Style Books Movies Technology Health Education Others Others |Music Movies Travel |World Music Movies Real Estate Health Travel Politics Economy Others Others |Others |Art & Design World Sports Fashion & Style Science Health Education Others Others |Others |Art & Design Sports Fashion & Style Science Food Health Travel Education Economy Others Others World |World Fashion & Style Health Travel Others |Fashion & Style |Sports Fashion & Style Music Technology Health Education Others Others Art & Design World Technology Health Others World Others Others |Sports |World Sports Technology Science Health Education Others Others Music |Sports Fashion & Style Music Technology Science Food Real Estate Health Others |Education |Technology Science Health Education Others |Others |Fashion & Style |World Technology Health Education Others Others Art & Design Sports Fashion & Style Technology Health Education Others Others Others Music |Sports Fashion & Style Music Technology Science Health Education Others Others |World Fashion & Style Health Education Others World Others |Sports Fashion & Style Music Health Education Others Others Others |World Sports Economy |Sports Fashion & Style Technology Health Education Economy Others Others |World Sports Fashion & Style Technology Science Health Education Others Others World only showing top 20 rows

Fraining 4 MulticlassPerceptron models using 57 training data.....

Fraining 4 MulticlassPerceptron models using 57 training data.....

redicting labels of test Tweets after trained 57 Category Feedback Tweets:

\veragePerceptron	NewsPerceptronIgnoreWeight	NewsPerceptronStoreWeight	OnlinePerceptro
)thers	Health	Art & Design World Sports Fashion & Style Technology Science Real Estate Health Education	Others
)thers	Fashion & Style Travel	World Sports Fashion & Style Movies Technology Real Estate Health Education	Others
)thers	Science Food	Art & Design Sports Fashion & Style Technology Science Health Education	Others
)thers	Sports	Art & Design World Sports Fashion & Style Books Music Movies Technology Science Health Education	Others
thers	Music Movies Travel	World Sports Music Movies Technology Real Estate Health Travel Education Politics	Others
thers	Others	Art & Design World Sports Fashion & Style Technology Science Health Education	Others
thers	Others	Art & Design World Sports Fashion & Style Science Food Real Estate Health Travel Education Your Money Economy	Others
thers	World Fashion & Style Travel	World Sports Fashion & Style Technology Real Estate Health Travel Education	Others
thers	Fashion & Style	World Sports Fashion & Style Books Music Technology Science Food Health Education	Others
thers	World	Art & Design World Sports Fashion & Style Books Music Technology Health Education	Others
thers	Sports Health	World Sports Fashion & Style Technology Science Health Education	Others
thers	Music	Sports Fashion & Style Books Music Technology Science Real Estate Health Education	Others
thers	Science Education	Sports Books Technology Science Health Education	Others
thers	Fashion & Style	World Sports Music Technology Real Estate Health Education Your Money	Others
thers	Others	Art & Design World Sports Fashion & Style Music Technology Science Health Education	Others
thers	Music	Sports Fashion & Style Books Music Technology Science Health Education	Others
thers	World	Art & Design World Sports Fashion & Style Music Technology Health Education	Others
thers	Others	Art & Design World Sports Fashion & Style Music Technology Health Education	Others
thers	World Sports Fashion & Style Economy	Sports Fashion & Style Music Technology Health Education Your Money Economy	Others
thers	World Health	World Sports Fashion & Style Music Technology Science Health Education	Others

nly showing top 20 rows

## Compare 4 MulticlassPerceptron Models with different iterations and memory state

bin/spark-submit category\_tweets\_perceptron.py --master local[10]

Training data is Feedback of correct categories of tweets (400)

Predict categories of test tweets using models after online trained every 100 tweets

- OnlinePerceptron model with weight initialized as zeros, iteration=10
- OnlinePerceptron model with weight initialized as trained model by news(ignored history of News), iteration=10
- AveragePerceptron model with weight initialized as trained model by news(ignore history of News), iteration=1
- AveragePerceptron model with weight initialized as trained model by news(ignore history of News),
   iteration=10

```
print "Training 4 MulticlassPerceptron models using",dataset.count(),"training data.......\n"
# train an Online Perceptron model(Iteration=10) use part of training data, initialize model with all zeros
initialModelOP.train(traindata,trainlabels,method="Online",MaxItr=10)
# train an Online Perceptron model(Iteration=10) use part of training data, initialize model with news trained model(ignore history)
initialModelAP.train(traindata,trainlabels,method="Online",MaxItr=10)
# train an Average perceptron model(Iteration=1) use part of training data, initialize using News trained Model(ignore history)
loadModelOP.train(traindata,trainlabels,method="Average",MaxItr=1)
# train an Average perceptron model(Iteration=10) use part of training data, initialize using News trained Model(ignore history)
loadModelAP.train(traindata,trainlabels,method="Average",MaxItr=10)

print "Predicting labels of test Tweets after trained ",dataset.count()," Category Feedback Tweets:\n"
```

category	text
)thers	really hate folks w. pissy ass attitudes for nun shit no where near cute
Others	" even ignoring @RuPaul in her quest to become a mindless mogul." Never! https://t.co/qr4mecHmSv
Food	Bout to eat at CB 👽 Love their breakfast
Others	I hate my friends lmao https://t.co/Esb04AktBg
Your Money	Every tax dollar invested in family-planning saves taxpayers 7 in obvious, predictable costs.
Others	dead week, man. It gets people. https://t.co/C8zislnRa3
Others	"Where dreams are made of" is not poetic license, it's just bad grammar.
Politics	Soldiers dying in war with naksals, party ministers and people busy celebrating delhi MCD election, not good!
Others	Bad little bitch, wanna know how her lips taste
Travel	We have these in downtown Omaha, Nebraska ⊕https://t.co/hvUg9PoqGM
Others	It's weird that you can remember dreams, that make no sense. But for some reason it will stick with you.
Fashion & Sty	e @Phil_Lewis_ Next up, underwear with artificial skid marks.
Others	@clairemacckyd Are you referring to the dinked middle? Do you have 7" "spider" adapter? Or did you mean something e https://t.co/2bCHaynfsF
	e If u put actual clothes on & do your makeup everyday before class, what r u doing?!?!?
Others	@Haylie_Evald3 so jealous ♥
Others	So tired of everybody & everything.
World	China Launches Aircraft Carrier, Boosting Military Presence
Others	motherfucking sucks.
Television	these men really are working hard for their spinoff. https://t.co/0cHaw10UEt
Health	Hospital againnnnn 🚛

+	+	.+
+	NewsPerceptronAPItr10	NewsPerceptronOPItr10
Sports Fashion & Style Health Travel  Others	Sports Fashion & Style Health Travel	Sports Fashion & Style Health Travel
Fashion & Style Movies Health Travel	Fashion & Style Movies Health Travel	Fashion & Style Movies Health Travel
Sports Fashion & Style Technology Science Health Education  Art & Design Sports Health Education	Sports Fashion & Style Technology Science Health Education  Art & Design Sports Health Education	Sports Fashion & Style Technology Science Health Education  Art & Design Sports Health Education
Others  Music Movies Real Estate Travel	Music Movies Real Estate Travel	Music Movies Real Estate Travel
Others  World Health	World Health	World Health
Others  Art & Design Fashion & Style Food Health Travel Your Money Economy	y Art & Design Fashion & Style Food Health Travel Your Money Econom	y Art & Design Fashion & Style Food Health Travel Your Money Economy
Others		
World Fashion & Style Travel   10thers	World Fashion & Style Travel	World Fashion & Style Travel
Sports Fashion & Style Music	Sports Fashion & Style Music	Sports Fashion & Style Music
Others   World Health	World Health	World Health
Others  Sports Science Health	Sports Science Health	Sports Science Health
Others  Fashion & Style Music Real Estate	Fashion & Style Music Real Estate	Fashion & Style Music Real Estate
Others  Science Education	Science Education	Science Education
Others  Technology Health Education Your Money	Technology Health Education Your Money	Technology Health Education Your Money
Others  Fashion & Style Education	Fashion & Style Education	Fashion & Style Education
Others  Fashion & Style Music Science Health	Fashion & Style Music Science Health	Fashion & Style Music Science Health
Others    World	World	World
Others	Others	Others
Others    World Sports Fashion & Style Health Your Money Economy	World Sports Fashion & Style Health Your Money Economy	
Others	Two to sports resirron a style hearth roun noney economy	worth sports rashron a style hearth roun noney economy
Vorld Sports Health  Others	World Sports Health	World Sports Health
+	+	+
tt		
only showing top 20 rows		

Training 4 MulticlassPerceptron models using 78 training data.....

Predicting labels of test Tweets after trained 78 Category Feedback Tweets:

+		+	<b>+</b>
NewsPerceptronAPItr1	NewsPerceptronAPItr10	NewsPerceptronOPItr10 	OnlinePerceptron
Others  Fashion & Style Movies Travel  Technology Science  Sports  Music Movies Travel  World  Art & Design Fashion & Style Food Travel Economy  World Fashion & Style  Fashion & Style  Sports  Music	Others Fashion & Style Movies Food Travel Science Food Others Music Movies Travel Economy World Art & Design Fashion & Style Food Travel Economy		Others Others Others Others Others Others Others
Music  Vorld  Others		Fashion & Style  Music  World  Others  World Fashion & Style Economy  World	Others Others Others Others Others Others

only showing top 20 rows

Training 4 MulticlassPerceptron models using 75 training data.....

Training 4 MulticlassPerceptron models using 75 training data.....

Predicting labels of test Tweets after trained 75 Category Feedback Tweets:

NewsPerceptronAPItr1	NewsPerceptronAPItr10	NewsPerceptronOPItr10	OnlinePerceptror
Others	Others	Politics	Others
Fashion & Style Travel	Fashion & Style	Others	Others
Science	Food	Food	Others
Sports Health	Others	Others	Others
Music Movies Travel	Music Movies Travel Economy	Movies	Others
World	World	World	Others
Food Travel Economy	Economy	Others	Others
World Fashion & Style Travel	World Fashion & Style Travel	World	Others
Fashion & Style	Fashion & Style	Fashion & Style	Others
World	World	World	Others
Others	Others	Others	Others
Music	Music	Music	Others
Education	Education	Education	Others
Technology	Technology	Fashion & Style	Others
Fashion & Style	Fashion & Style	Others	Others
Music	Music	Others	Others
World	World	World	Others
Others	Others	Others	Others
World Sports Fashion & Style Eco	nomy World Fashion & Style Economy	World Fashion & Style Econo	omy Others
World	Others	Health	Others

only showing top 20 rows

Training 4 MulticlassPerceptron models using 79 training data.....

Fraining 4 MulticlassPerceptron models using 79 training data.....

redicting labels of test Tweets after trained 79 Category Feedback Tweets:

NewsPerceptronAPItr1	NewsPerceptronAPItr10	NewsPerceptronOPItr10	OnlinePerceptron
Health	+	+  Politics	+  Others
Fashion & Style Movies Travel	Fashion & Style Travel	10thers	Others
	Science	Technology	
Sports Technology Science			Others
Sports Health Education	Others	Others	Others
Music Movies Travel	Music Movies Travel Economy	Music Movies Travel	Others
World	World	World	Others
Art & Design Fashion & Style Food Health Travel Your Money Economy	Art & Design Fashion & Style Food Travel Economy	Others	Others
World Fashion & Style Travel	World Fashion & Style Travel	World Travel	Others
Fashion & Style Music	Fashion & Style	Fashion & Style	Others
World Health	World	World	Others
Sports Health	Others	Others	Others
Music	Music	Fashion & Style	Others
Science Education	Education	Others	Others
Technology Education Your Money	Technology	Others	Others
Fashion & Style	Fashion & Style	Others	Others
Music	Music	Others	Others
World	World	World	Others
Others	10thers	Others	Others
World Sports Fashion & Style Health Your Money Economy	World Fashion & Style Economy	Fashion & Style Economy	
World Sports	Others	Health	Others

only showing top 20 rows

Training 4 MulticlassPerceptron models using 82 training data.....

Training 4 MulticlassPerceptron models using 82 training data.....

Predicting labels of test Tweets after trained 82 Category Feedback Tweets:

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NewsPerceptronAPItr1	NewsPerceptronAPItr10	NewsPerceptronOPItr10	0 OnlinePerceptro
Sports Fashion & Style Real Estate Health Your Money	Others	Others	Others
World Fashion & Style Movies Health Travel	Fashion & Style Movies Food Travel	Others	Others
Sports Fashion & Style Technology Science	Science Food	Technology Food	Others
Sports Health Education	Sports	Others	Others
Music Movies Travel Economy	Movies Economy	Your Money	Others
World	[World	Others	Others
Art & Design Sports Fashion & Style Music Food Health Travel Your Money Economy	Art & Design Fashion & Style Food Travel Economy	Others	Others
World Fashion & Style	World Fashion & Style	Politics	Others
Sports Fashion & Style Music Your Money	Fashion & Style	Others	Others
World Health	World	Travel	Others
Sports Health	Sports	Others	Others
Fashion & Style Music	Music	Fashion & Style	Others
Technology Science Education	Food Education	Others	Others
Technology Health Education Your Money	Technology	Fashion & Style	Others
Sports Fashion & Style Technology Education	Fashion & Style	Others	Others
Fashion & Style Music Technology Your Money	Music	Others	Others
World	World	World	Others
Sports	Others	Others	Others
World Sports Fashion & Style Music Health Your Money Economy	  World Fashion & Style Economy	Television	Others
World Sports Health	World	Health	Others
aly showing top 20 rows		ł	·+

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