

DATA DRIVEN

TOYOTA Customer 360° on
Apache Spark™

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Toyota Motor Sales IT Research and Development
Final

TOYOTA Big Data History

2015



C360 - Next Gen Insights Platform
Over 6B Records

2014



C360 - Customer Experience Analytics
Over 700M Records

2013



C360 - Toyota Social Media Intelligence Center
Over 500M Records

2012



Product Quality Analytics v2
Over 120M Records

2011



Marketing and Incentives Analytics
70M Records

2010



Product Quality Analytics
Over 60M Records



 TEAM TOYOTA

R&D
Data Engineering
Infrastructure
Enterprise Architecture

SPARK SUMMIT 2014

現地現物

Genchi Genbutsu
“Go Look, Go See”





- Compute
- Streaming
- Machine Learning

ACTIONABLE INSIGHTS



Data Engineering

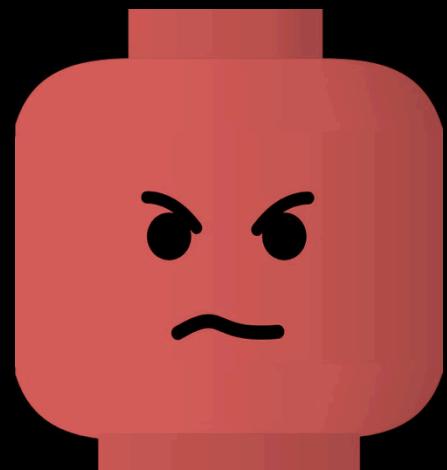
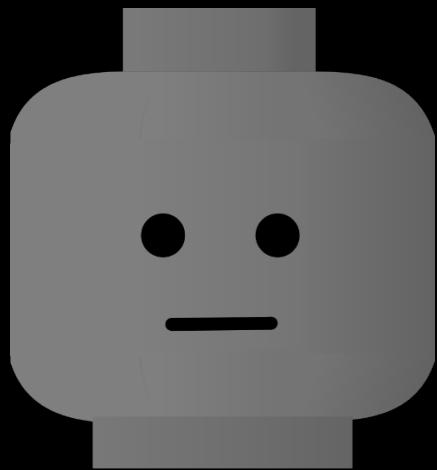
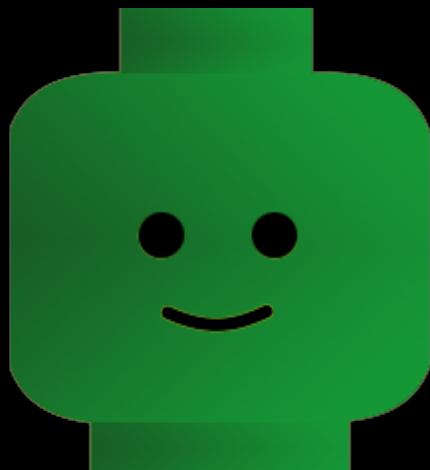
Customer Experience **original** Batch Job

160 hours (6.6 days)

Same job re-written using **Apache Spark** ...

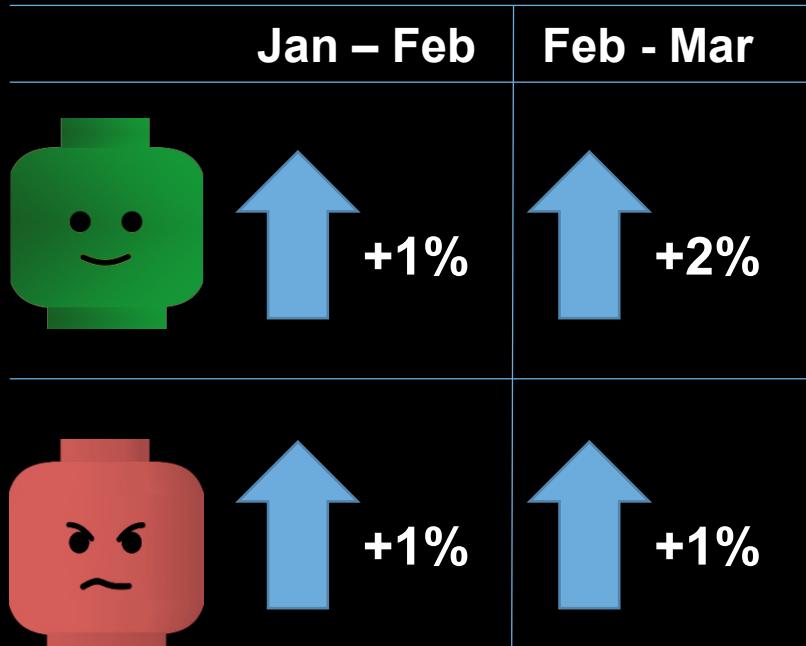
4 hours

Existing Tools



Existing Tools

Toyota Social Opinion 2013



Toyota Online Conversations by the Numbers



2014
Study

- 40%** Retailers Selling Toyota Vehicles
- 11%** Opinions on Marketing Campaigns
- 10%** Feedback on Dealer Sales and Service Experiences
- 9%** Opinions on Product Styling and Features
- 8%** People In Market for a Toyota
- 8%** Incident Reports Involving a Toyota Vehicle
- 7%** Feedback on Product Quality
- 5%** Customers Advocating for the Brand
- 2%** Completely Irrelevant



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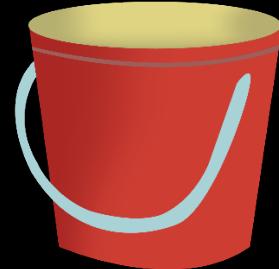
50%
Noise





Categorize and Prioritize incoming Social Media interactions in Real-Time using Spark MLLib

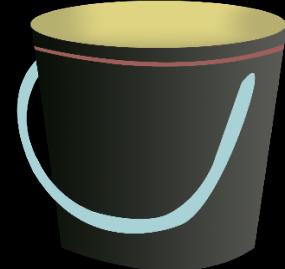
Campaign
Opinions



Customer
Feedback

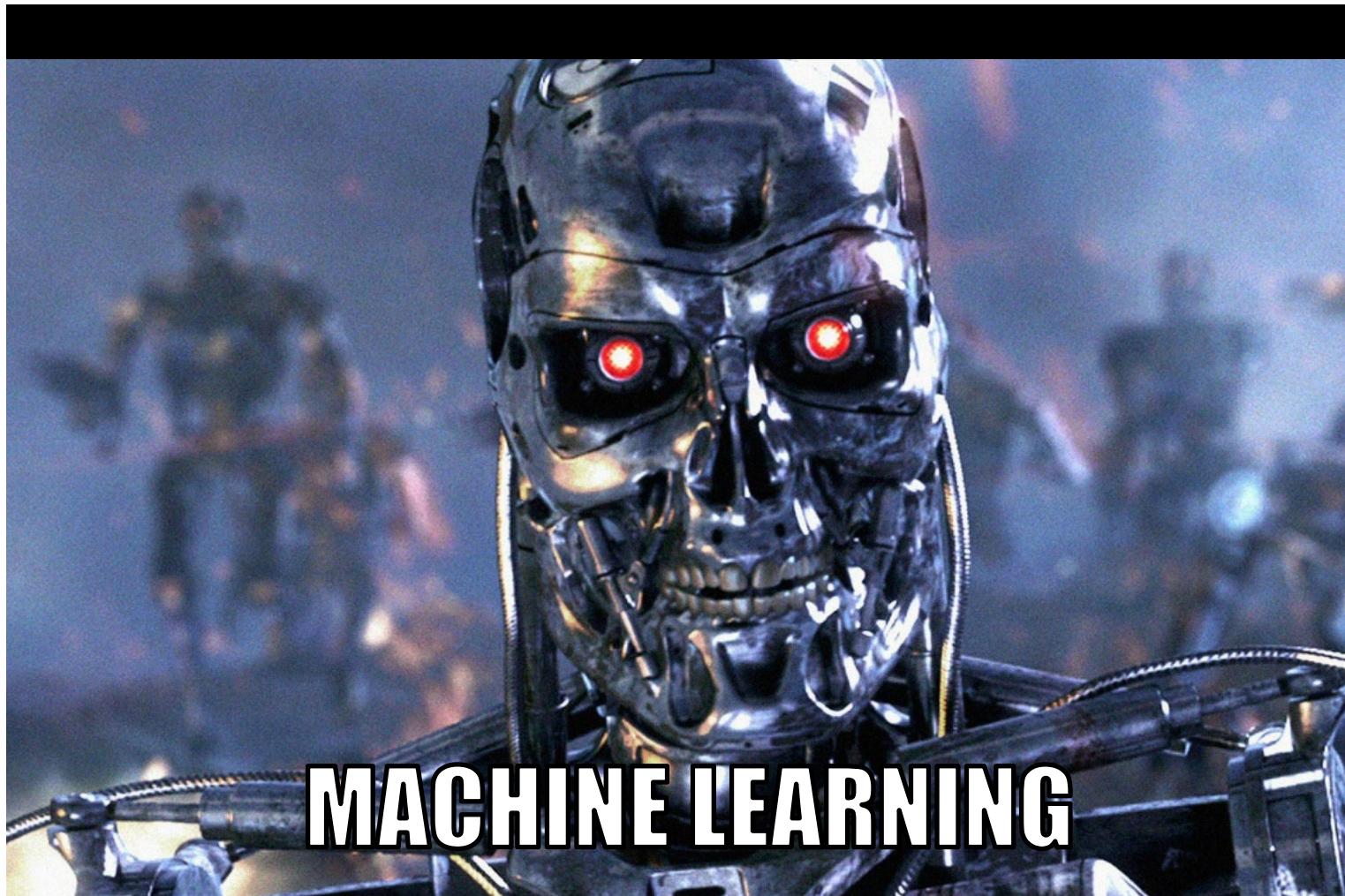


Product
Feedback



Noise





First Spark MLlib Experiment

Time-box project to 12 Weeks
Classify at min 80% accuracy

- Seat Cover Wrinkles/Cracking
- **Brake Noise**
- Shift Quality
- Oil Leaks
- HVAC Odor
- Dead Battery
- Rodent Wire Harness Damage
- Paint Chips



**Describe this
issue...**



Brakes

Noise





When I'm backing up in my 2012 Prius, it sounds like something hanging up or scraping as it rotates and only happens in the morning..



I hear a squeak coming from the back wheel of my Prius as I pull out from my driveway in the morning.



Identify Training Data for Brake Noise Model

BRAKES

Mark the box next to the problem

- 01) Brakes pull to the left
- 02) Brakes pull to the right
- 03) Brake noise (please specify front/rear/both)
- 04) Brake vibration (please specify front/rear/both)
- 05) Brake application/pedal feel (soft/hard)
- 06) Braking effectiveness
- 07) Parking brake
- 08) ABS Brakes (ABS light/ABS system)
- 09) VSC (VSC light/VSC system)
- 10) Brake problem other than listed above

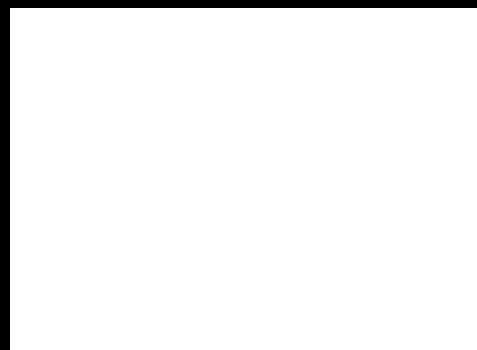
When, Where, What

When it's cold
the brakes make
noise specifically
when backing up
they squeak

Problem Severity (Check One)

Not At All Serious 1 2 3 4 5 Extremely Serious

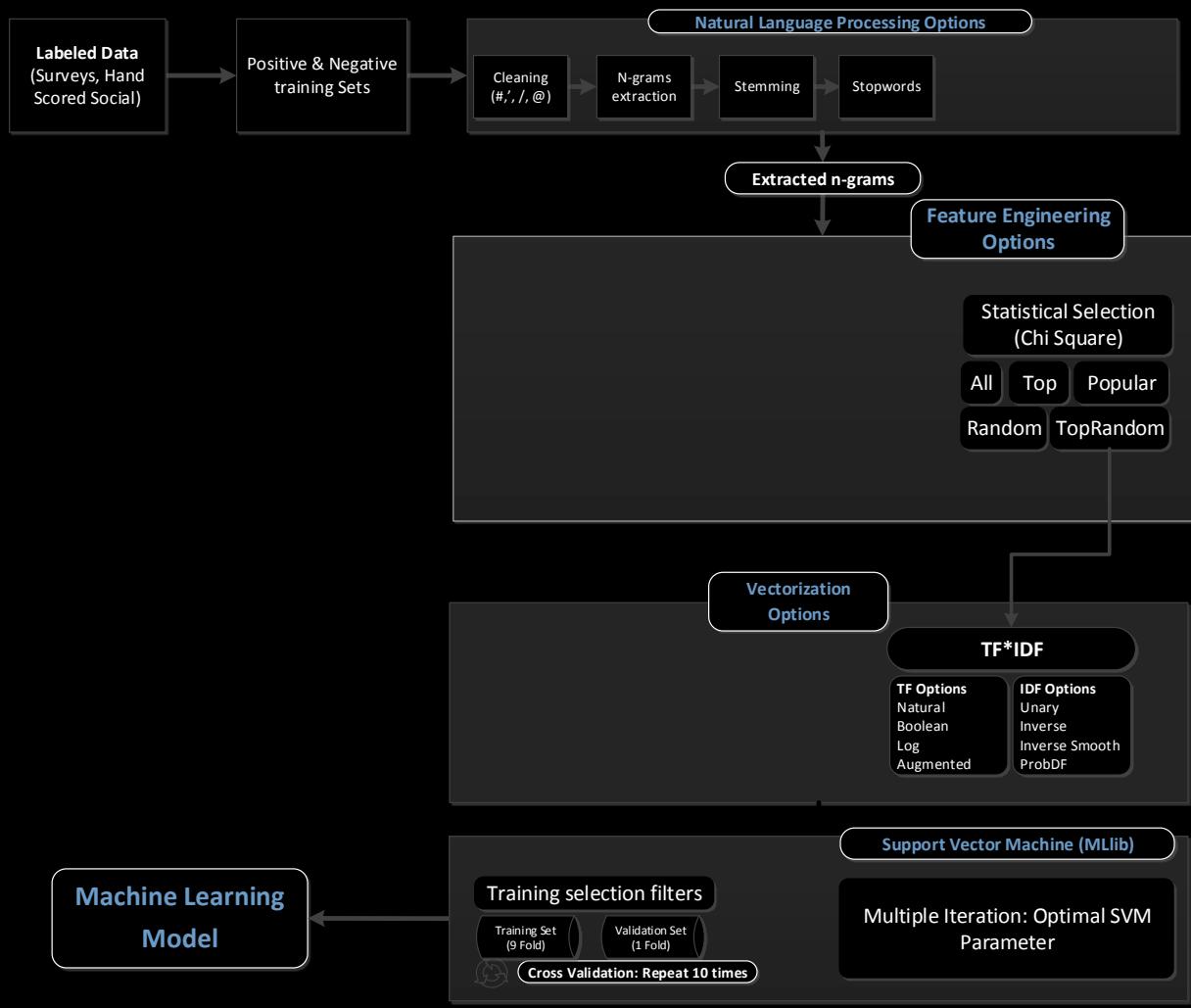
Extract Category (Label) matching Model Objective



Extract features from the Verbatim Text



Social ML Pipeline



Extract Text Features

```
object ChiSquareFeatureSelector {

  def selectFeatures(documents: RDD[LabeledNgramInfos], chiSquareCriticalValue: Double, rareTermThreshold: Int = 3): ChiSquareFeatureSelector = {

    val corpusSize = documents.count

    val classCountPerCorpus = countDocumentsByClassification(documents)

    val featuresRDD = documents
      .flatMap{ doc =>
        doc.features.map { ngram => (NgramClassification(ngram.term, doc.label), 1) }
      }
      .reduceByKey(_ + _)
      .map( x => (x._1.ngram, ClassificationCount(classificationKey=x._1.classificationKey, count=x._2) ) )
      .groupByKey()
      .flatMap { case(ngram, classificationCounts) =>
        calculateChiSquare(ngram, classificationCounts, classCountPerCorpus, corpusSize)
      }
      .filter(includeThisFeature(_, chiSquareCriticalValue, rareTermThreshold))
      .cache()

    new ChiSquareFeatureSelector(featuresRDD)
  }
}
```



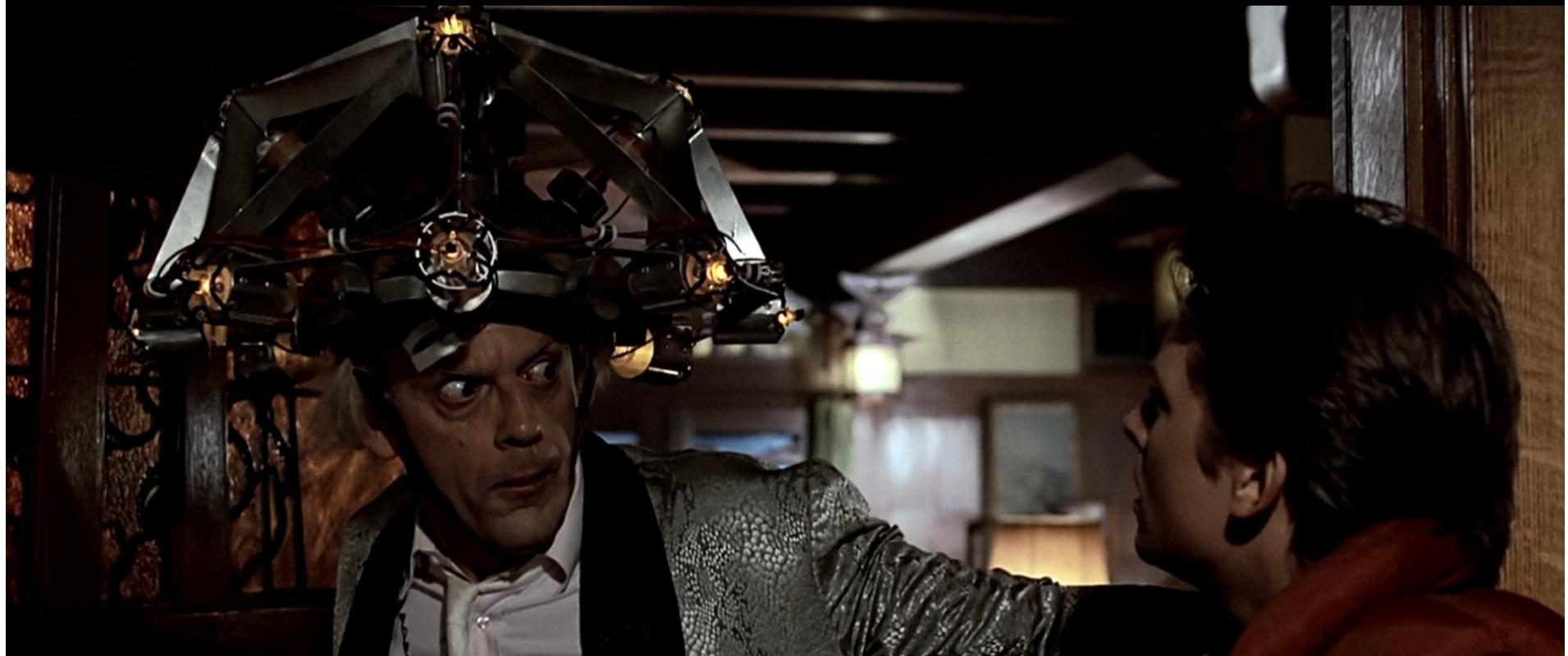
Train Predictive Model

```
object KFoldsTrainingModel {  
  def train(  
    labelName: String,  
    kFolds: Int = 10,  
    numSvmIterations: Integer = 10,  
    svmGamma: Double = 0.5,  
    svmRegParam: Double = 1.0,  
    kernelType: String = "linear",  
    negDocVectors: RDD[LabeledPoint],  
    posDocVectors: RDD[LabeledPoint])(implicit sc: SparkContext) : KFoldsTrainingModel = {  
  
  val votesRequired = math.ceil((kFolds.toDouble - 1) / 2)  
  val foldsArray = Array.fill(kFolds)(1.0 / kFolds)  
  
  val splits0 = negDocVectors  
    .randomSplit(foldsArray, seed = 11L)  
  
  val splits1 = posDocVectors  
    .randomSplit(foldsArray, seed = 11L)  
  
  val mlModels = kernelType match {  
    case "linear" => trainKFoldsLinear(splits0, splits1, kFolds, numSvmIterations, svmRegParam)  
  }  
  
  // run predictions for each of the k-folds models  
  val testFold = kFolds - 1  
  val testData: RDD[LabeledPoint] = splits0(testFold).union( splits1(testFold) ).cache()  
  val testModels = mlModels.take(kFolds - 1)  
  
  kernelType match {  
    case "linear" => {  
      val predictParams = Seq(0.0)  
      for( predictParam <- predictParams ) {  
        predictAll(labelName, testModels, testData, votesRequired, predictParam.asInstanceOf[Double])  
      }  
    }  
  
    case _ => predictAll(labelName, testModels, testData, votesRequired)  
  }  
  
  new KFoldsTrainingModel(  
    labelName,  
    mlModels.toList,  
    votesRequired)  
}
```

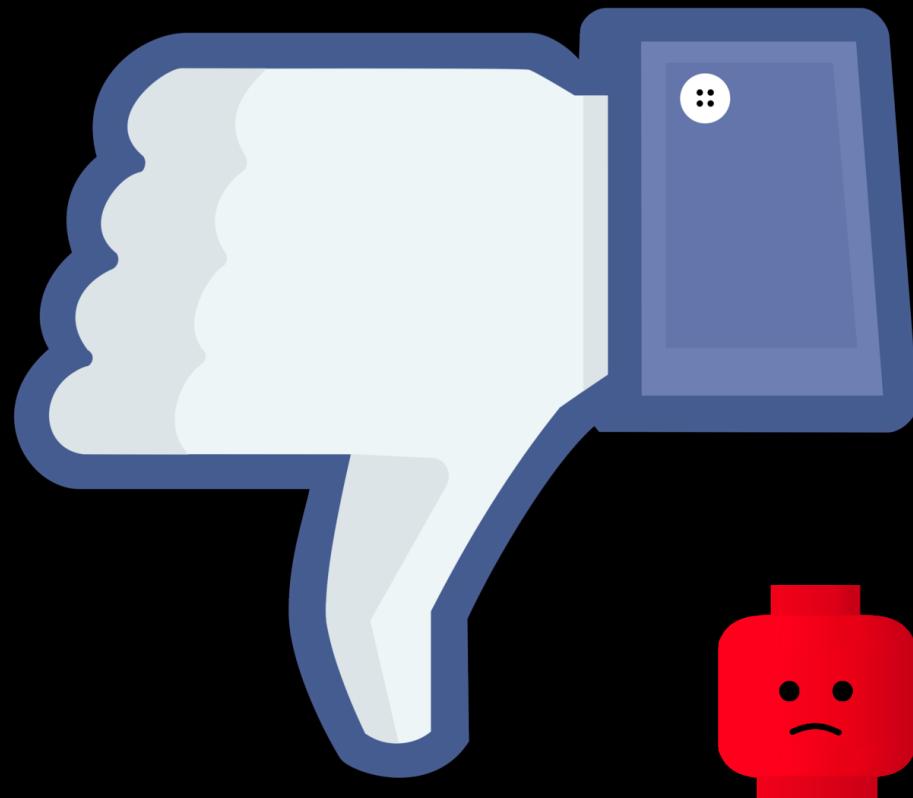


TOYOTA

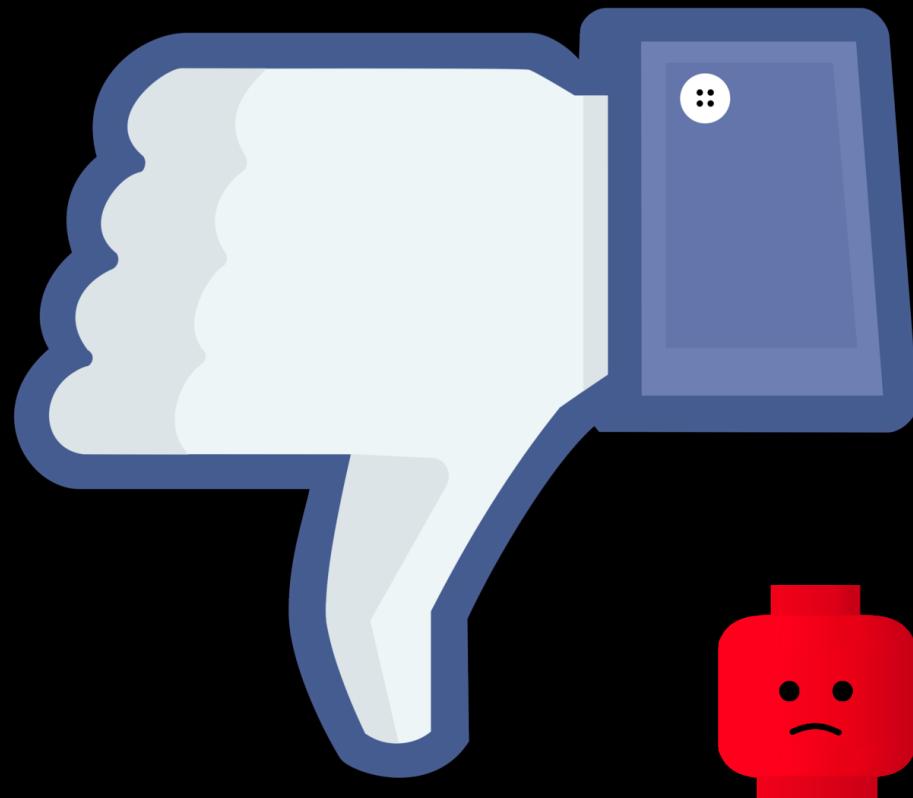
```
private def predictAll(  
  LabelName: String,  
  mLModels: ParSeq[ClassificationModel],  
  testData: RDD[LabeledPoint],  
  votesRequired: Double,
```



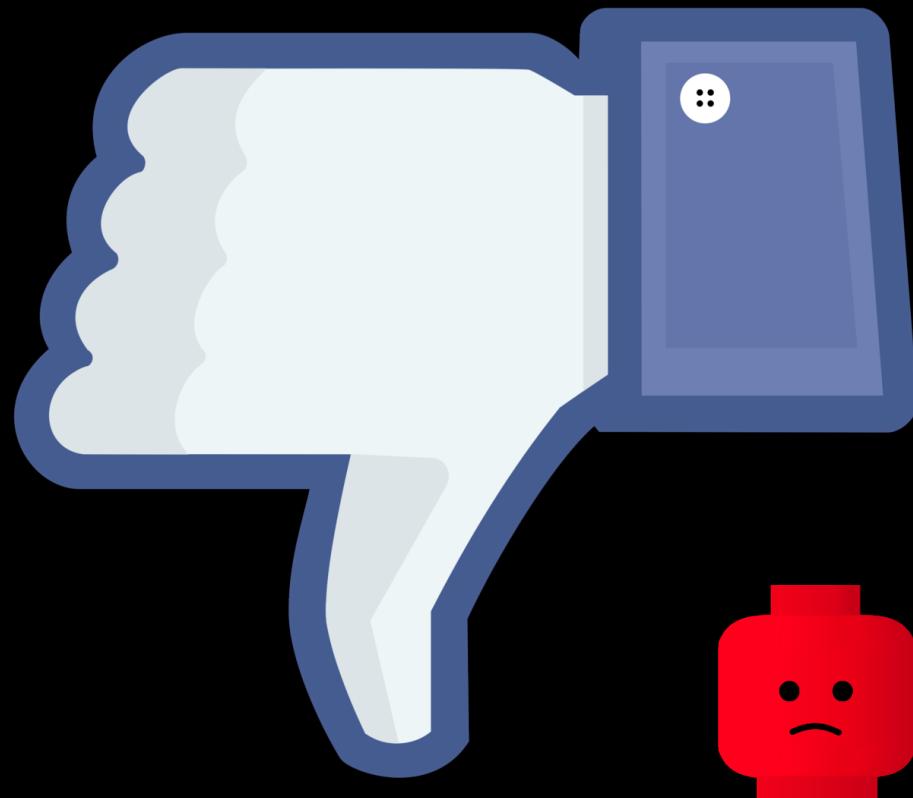
Ver 1
56%
Accuracy



Ver 3
36%
Accuracy



Ver 8
35%
Accuracy



False Positives



I just had my friend at the toyota dealer rotate my tires and he said ... that the **brake** pads are getting thin really fast. So what should I do when they get too thin in the future and start to **squeak**?



False Positives



i cut the iac hose as shown in figure 20 in the manual but when i start the car, it started gasping for air... **choking...**
sounds like it's about to die out.

i bought the power **brake** check valve (80190 part for kragen)... but either i'm not installing it right or it's the wrong size... i have no idea.





Explicit Semantic Analysis

WIKIPEDIA
The Free Encyclopedia

Article Talk Read Edit View history Search

Drum brake

From Wikipedia, the free encyclopedia

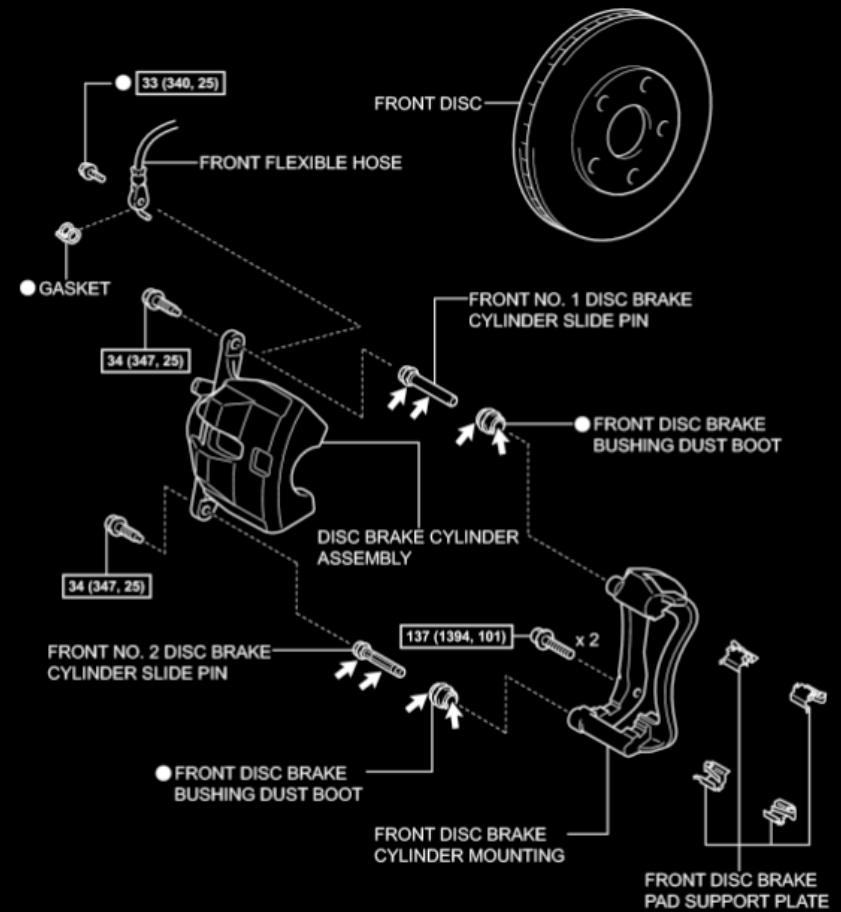
This article relies largely or entirely upon a single source. Relevant discussion may be found on the talk page. Please help improve this article by introducing citations to additional sources. (July 2008)

A **drum brake** is a brake that uses friction caused by a set of shoes or pads that press against a rotating drum-shaped part called a brake drum.

The term *drum brake* usually means a brake in which shoes press on the **inner surface** of the drum. When shoes press on the outside of the drum, it is usually called a *clasp brake*. Where the drum is pinched between two shoes, similar to a conventional *disc brake*, it is sometimes called a *pinch drum brake*, though such brakes are relatively rare. A related type called a *band brake* uses a flexible belt or



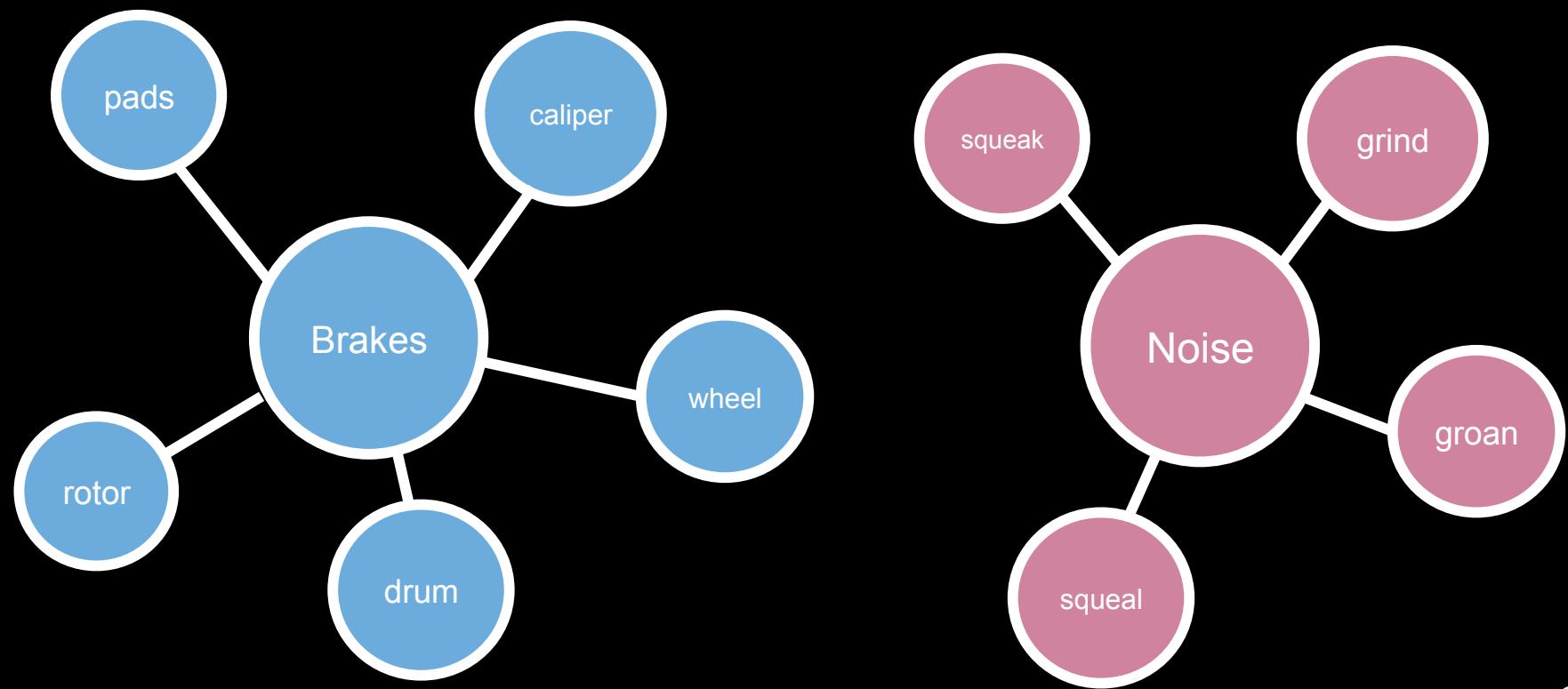
A drum brake with the drum removed, as used on the rear wheel of a car or truck. In this installation, a cable-operated parking brake operates the service shoes.



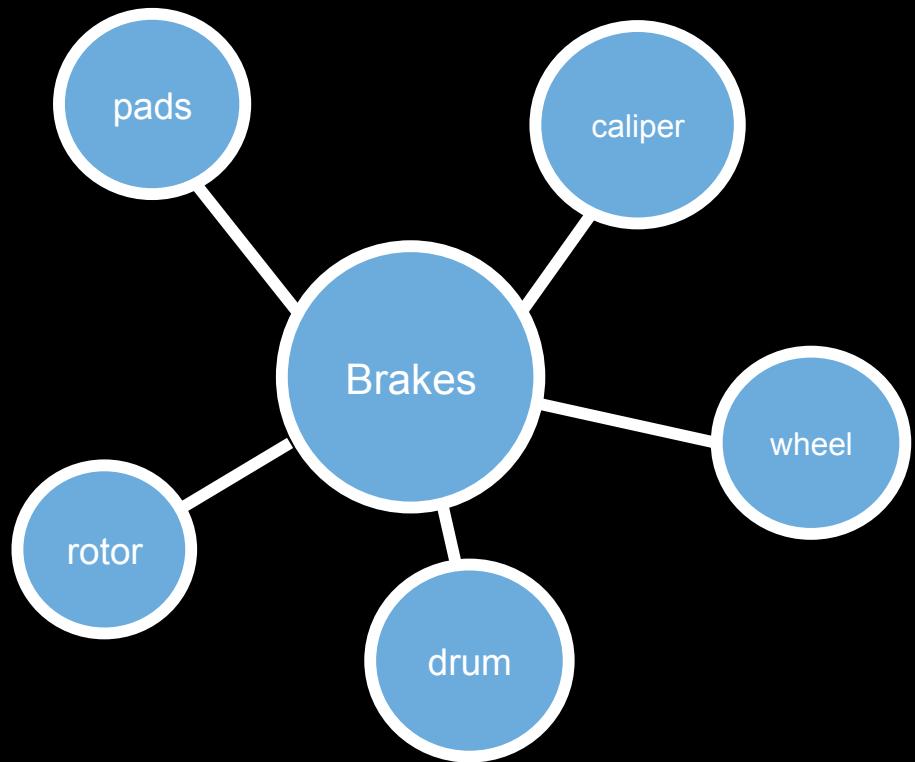
Brakes

Noise





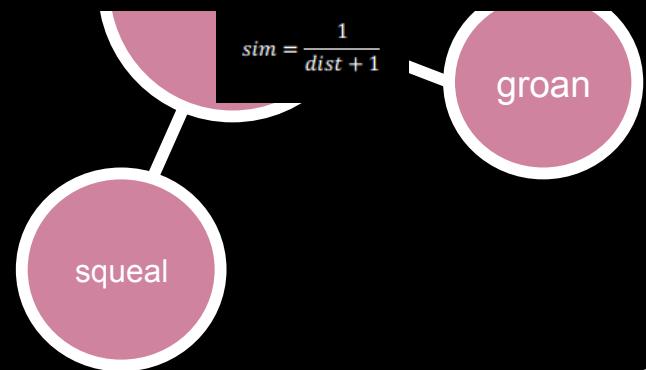
Distance Similarity Between Concepts



Distance is calculated between concepts based on the Minkowski distance formula.

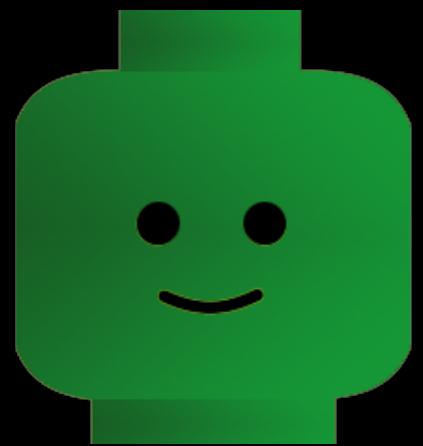
$$d(i, j) = \left(|x_{i1} - x_{j1}|^q + |x_{i2} - x_{j2}|^q + \dots + |x_{ip} - x_{jp}|^q \right)^{\frac{1}{q}}$$

$$sim = \frac{1}{dist + 1}$$



Ver 9
82%
Accuracy





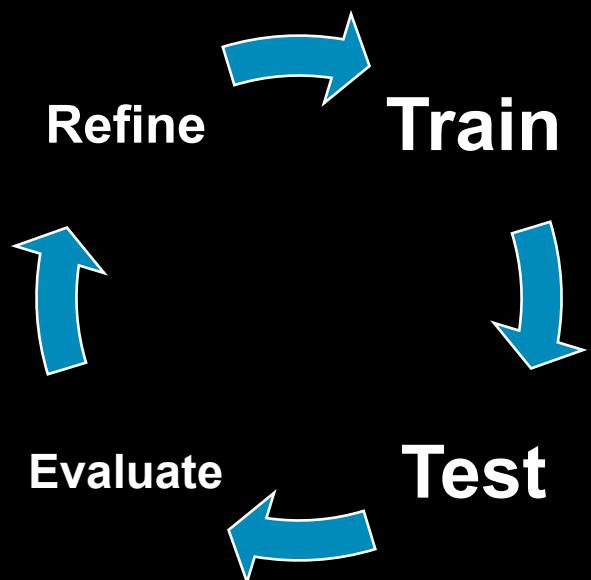
Kaizen

改善

= Continuous Improvement

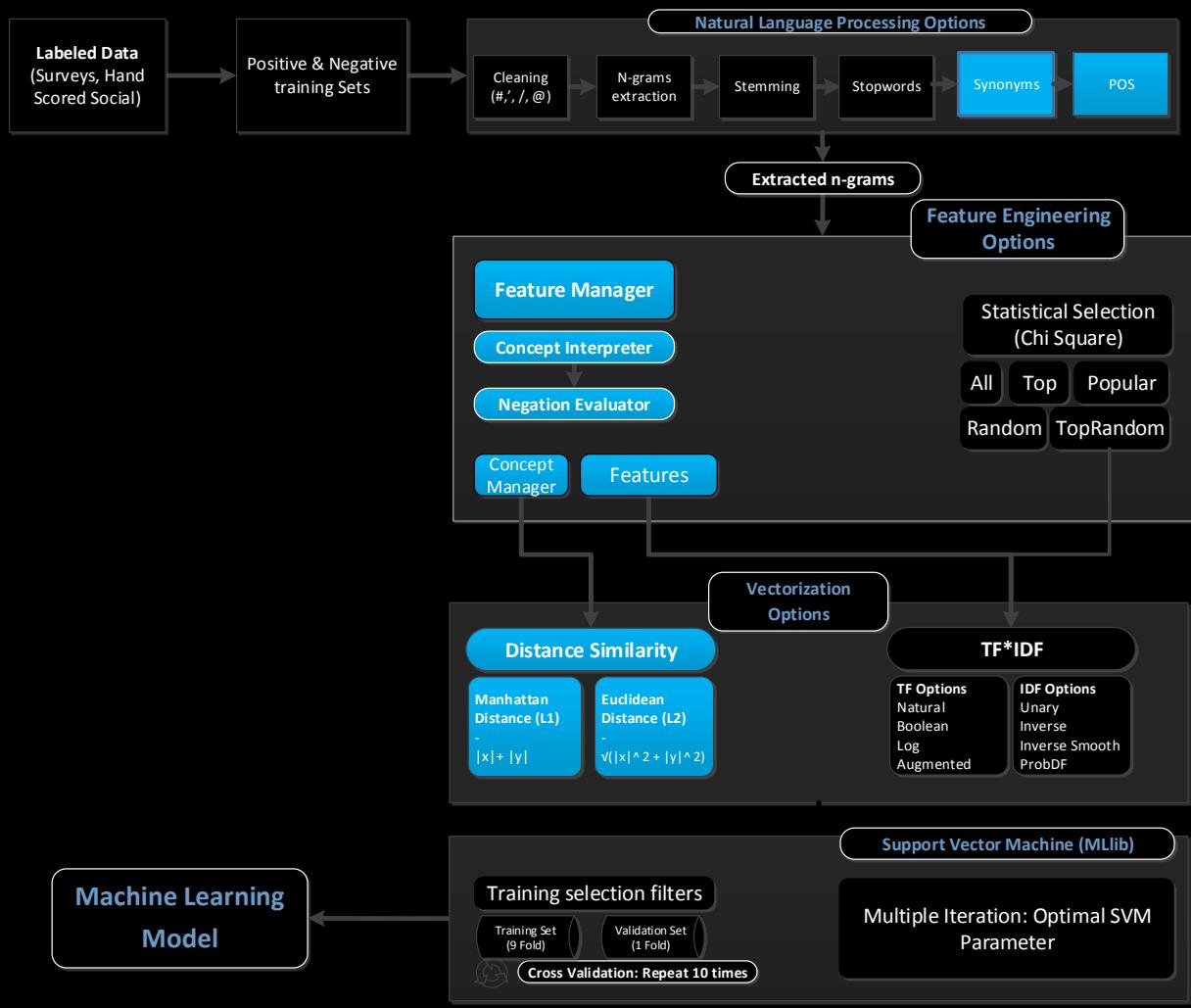


Kaizen





Social ML Pipeline



TODAY





FUTURE



Connected Vehicle Data



Consumer Data



Manufacturing Data



TEAM TOYOTA Spark Tips

- Education and Inclusion
- Pace Yourself
 - Design and Plan a Transitional Architecture to Incrementally Introduce Spark elements into your Applications
 - Use Joda Time for Date Comparisons

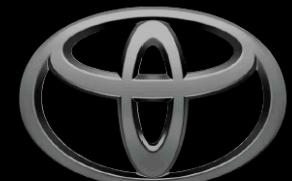




TEAM TOYOTA Lessons Learned

- Be mindful of AKKA versions when trying to Build a new Spark release to a packaged Hadoop Distribution
- Use SparkSQL versus DSLs for Joins
- Remember to configure Memory Fraction based on the size of your data.





*Let's
Go
Places*



Visit us at the **TOYOTA**
Booth here at the Spark
Summit Today.

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