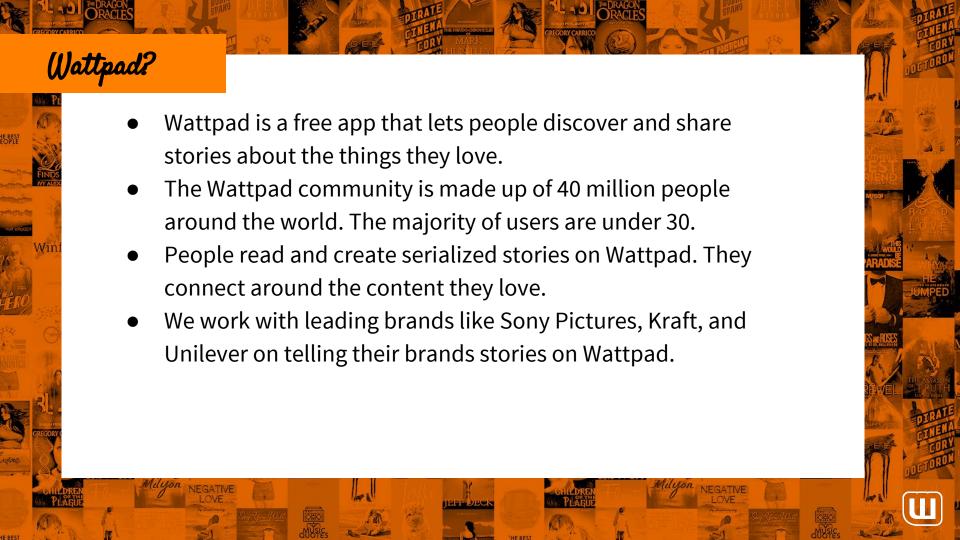
# (Welcome!

July 27th 2016





# Why do I care?

- Over 45M people use Wattpad each month
- 100k+ Signups per day
- Over 250,000,000 stories on the platform
- Spend 15.5 Billion (with a B) minutes
- Huge repository of data
- Let us tell you about how we use it all...

# Spark Stories

Wattpad

# Agenda

- First Spark project
- Running Spark on EMR
- Issues with Spark & Luigi
- Recommendations
- Doc2Vec & S3 Import
- Search

# First Project - Event Processing



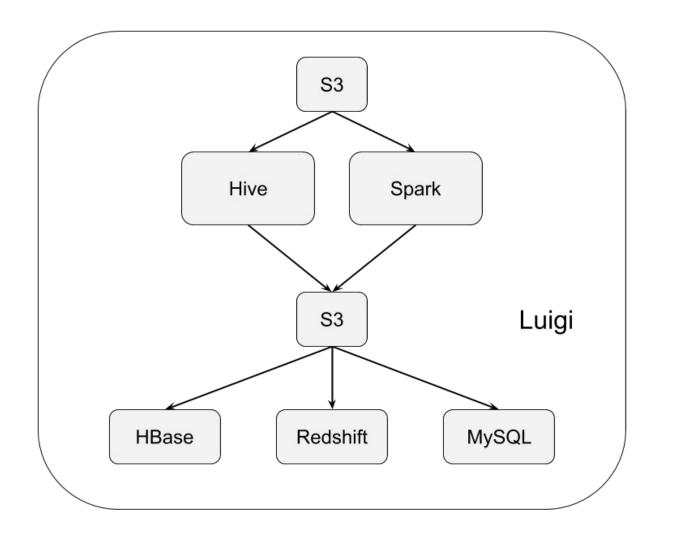
# First Project - Event Processing

- Originally using Shark & Spark 0.9
  - Did not perform well as number of events scaled
  - As JSON complexity increased, became difficult to parse in Hive
- Moved to Scala & Spark 1.1
- Currently on Spark 1.3
  - Accessing JSON schema had changed

# First Project - Event Processing

- 1.2B events per day
- Run on 3 r3.8xlarge instances < 3 hours

- Settings
  - SPARK\_NUM\_EXECUTORS ~ 90% of available CPUs
  - SPARK\_EXECUTOR\_MEMORY ~85% of available memory
    - NUM\_EXECUTORS \* EXECUTOR\_MEMORY



### **EMR**

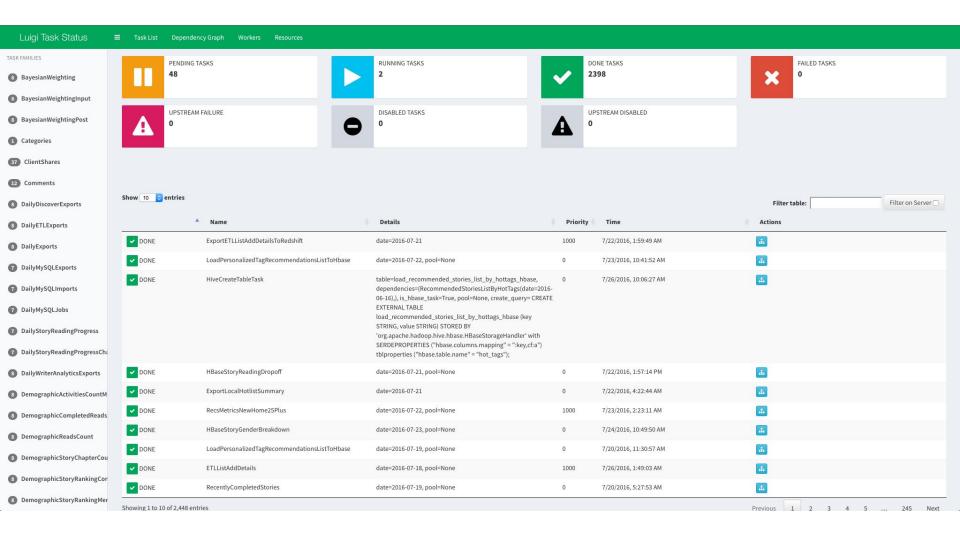
- Run new clusters every night
- No permanent cluster or metastore
  - Requires loading of schemas and data
- Spot instances
- Tied to Amazon's release cycle
  - Spark availability is tied to AMI version

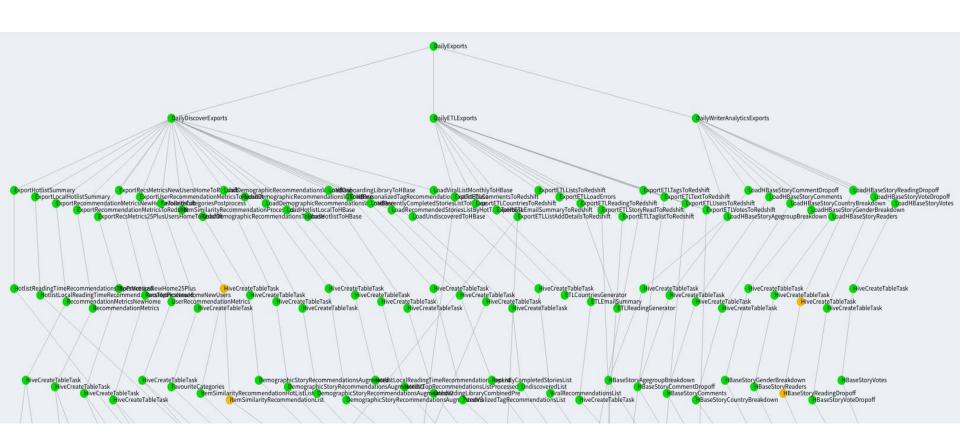
## Cluster settings

- NUM\_EXECUTORS ~ 20% of available cores
- EXECUTOR\_MEMORY ~ 13% of available memory
  - NUM\_EXECUTORS \* EXECUTOR\_MEMORY
- DEFAULT\_PARALLELISM = 3x NUM\_EXECUTORS

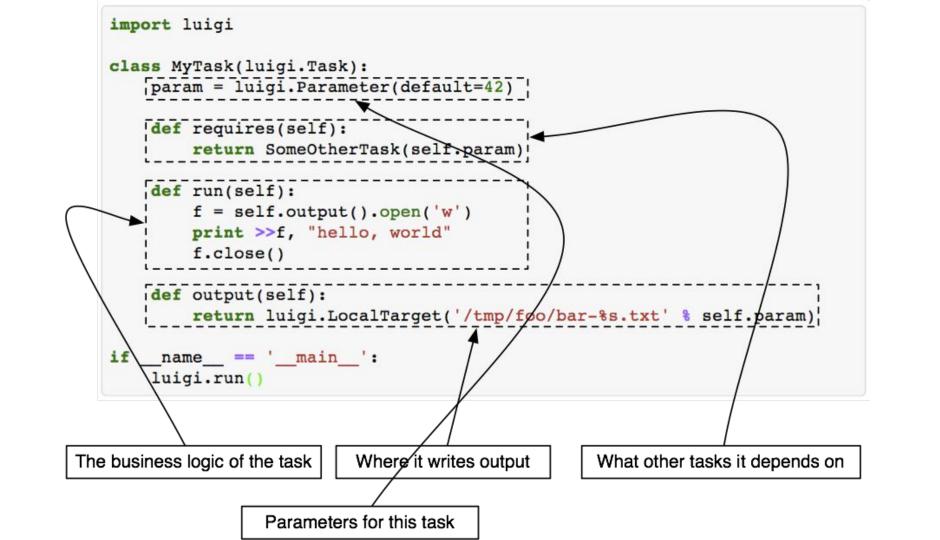
# Luigi

- Workflow Management Tool
  - Dependency resolution
  - Job scheduling
  - Visualization
- Written in python
- Community plugins





# Luigi Tasks



### Luigi - Spark Shared Context

- Included spark support starts & tears down the spark context for every task
  - Adds overhead
  - Duplicate work
  - Interim storage can be required

### How can we do better?

- Create our own Spark Task
- Keep a reference count of the number of tasks using the context
- Overwrite the callbacks

### Manipulate the context counter

```
def get context(self):
   if SharedSparkContext.applications == 0:
        self.initialize context()
    SharedSparkContext.applications += 1
    return self
def release context(self):
    SharedSparkContext.applications -= 1
   if SharedSparkContext.applications == 0:
        self.sc.stop()
```

### Create a singleton

```
class SharedSparkContext(object):
     metaclass = core.Singleton
   # Records the number of Spark applications currently using this Spark context
    applications = 0
   # Spark configuration and context information that will be shared across multiple
Spark applications
   conf = None
   sc = None
   sqlc = None
```

### Create our Luigi Spark Task

```
class BaseSparkApplication(luigi.Task):
   task namespace = 'spark'
    ctx = None
   def complete(self):
       is complete = super(BaseSparkApplication, self).complete()
        if not is complete and not self.ctx:
            self.ctx = SharedSparkContext().get context()
       return is complete
   def on success(self):
        self.ctx.release context()
   def on failure(self, exception):
        self.ctx.release context()
        return super(BaseSparkApplication, self).on failure(exception)
```

### Luigi - Spark UDFs

- Luigi code is only loaded onto the Master node
- Needed a way to make UDFs available to all nodes
- Extend our custom Spark Class

### On initialization

- Walk the UDF directory to get the files we need
- Zip the files
- Add the files to the Spark Context & register the UDFs

```
self.sc.addPyFile('file://' + zipfilename.filename)

# Create Hive context and register any UDFs available
self.sqlc = HiveContext(self.sc)
self._register_udfs(udfs)
```

```
def _register_udfs(self, udfs):
    # udfs are a list of named tuples ('SparkSQLUDF', 'name, func, return_type')
    for udf in udfs:
        self.sqlc.registerFunction(udf.name, udf.func, udf.return_type)
```

## Scala vs. Python

- Developers & Data Scientists tend to know Python better
- Integrates better with Luigi
- Better build process in Python
- Libraries not available in Scala
  - Nltk, scipy
- But scala is better sometimes
  - Event processing
  - Recommendations

### What's Next

- Open source
- Investigate SparkSession in Spark 2.0
- Investigate Tachyon

# Recommendations

### Recommendations

#### Recommendations



The Good Girt's Bad Boys: The HUMOR



Evolution (Book 1 News & Updates of POE



SCIENCE FICTION RANDOM ● 41.1M ★ 1.3M ● 1.1M ★ 29.3K ● 6.5M ★ 73.5K ● 21.7K ★ 1.6K ● 7.9K ★ 421



by Kelly Anne Bl....





WIN: The Atlantis



Grail (Book Thre ... by Stephen Clark... by VeraNazarian 

♣ Follow



#### Reading lists by Written In Action > Collections from a top profile



#### WIA Undiscovered 2



#### WIA Undiscovered



#### Stories by Pantopicon >



#### The Butcher's Daughter

O 197 ★ 21

The Butcher's Daughter follows 24 hours in the life of Cookie: club kid, mad genius, and controlled sociopath.



Air Thin and Eager

O 1.5K ★ 92

My name is Loretta but I prefer Lottle I'm closing in on my fifteenth year And if you think you have seen a pair of eyes more green Then you sure didn't see them around...



♣ Follow

#### Completed Stories



#Wattys2015



Evolution (Book 1 The Purge





(#Wattys2016)



Unbrokenworld: The Crystal Caves



Experiment

SCIENCE FICTION SCIENCE FICTION ⊕ 185K ★ 13,3K ⊕ 117K ★ 7.5K

#### #spock stories >



Star Trek Reader by Whoylan3135



Star Trek Reader Inserts 2 by Whoylan3135



Boarding



Shooting Star: Cassandra ( A Star Star Trek Oneby owenharper O 314K ★ 10.8K O 165K ★ 7.3K O 80.8K ★ 2.1K O 52.5K ★ 2K



Book 1: Love In The Stars (Star O 27.8K ★ 862 O 21.2K ★ 707



by CrazyDreamGlrl by UnaNova



#### #mystery stories >

Because you searched for wasted life



The Girl He Never Noticed

by makeandoffer @ 104M ★ 2.6M @ 28.4M ★



The Shy Girl Has a I'm a Model that's Struck (A Vampire Fatal Alliances Undercover as T... O 17.6M ★



by CaltSaral O 12.4M ★



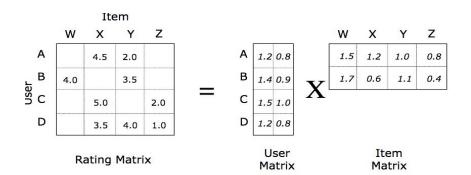


by CelineMahadeo by zabellerain O 10M ★ 361K O 12M ★ 319K

Marked by the

### Recommendations: MLlib

- ALS Matrix Factorization
  - Implicit signals
- Scala:
  - Performance
  - Library compatibility
- Personalized Recs:
  - o **800M** ratings
  - o 20M users x 10M stories
- Similar Items:
  - 1.3B ratings
  - 30M users x 25M stories



## Recommendations: Challenges

- 200 Trillion operations
  - Breeze (Runtime: 20+ hours)
  - JBLAS (Runtime: no change)
  - Native BLAS (Runtime 8 hours 60% ↓)
- Approximate Nearest Neighbors (Run time: 3 hours 62.5% )

- 1. Fast and Accurate Maximum Inner Product. Recommendations on Map-Reduce. Rob Hall and Josh Attenberg. Etsy Inc.
- 2. Improved Asymmetric Locality Sensitive Hashing (ALSH) for Maximum Inner Product Search (MIPS), Anshumali Shrivastava and Ping Li

### Recommendations: ANN With pySpark

- What is ANN?
  - searches subset of items
  - keeping accuracy as high as possible
- Spotify's Annoy<sup>1</sup> Library
  - speed
  - accuracy
  - python
- Run time: 40 mins (78% <sup>↓</sup>)
- Also available in Scala now<sup>2</sup>

```
# Add annoy index to spark
sc.addPyFile(os.path.join(base_path, annoy fname))
# KNN method
def get KNN(user features):
       t = AnnoyIndex(rank, metric='euclidean')
       t.load(SparkFiles.get(annoy fname))
       return [t.get_nns_by_vector(vector=features,
n=k, search k=-1, include distances=True)) for
user id, features in user features]
# Call to KNN method
recommendations = user features.mapPartitions
(get KNN)
```

### Recommendations: Diversification

- Balances accuracy and diversity of items in a list.
- Algorithms:
  - Pairwise IntraList diversification<sup>1</sup>
  - Clustering based genre diversification<sup>2</sup>
- Runtime:
  - 20 mins



- 1. The use of MMR, diversity-based reranking for reordering documents and producing summaries, Jaime Carbonell & Jade Goldstein
- 2. Improving recommendation lists through topic diversification, Cai-Nicolas Ziegler et al

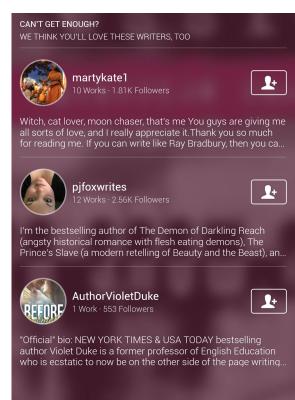
# Doc2Vec & S3 Import

### Doc2Vec

- A neural net technique for vectorizing segments of text.
   Extension of Word2Vec.
- Useful for performing content-based tasks: recommendations, clustering. Especially useful for cases where little user activity is available.



Use case: Content based user recommendation.



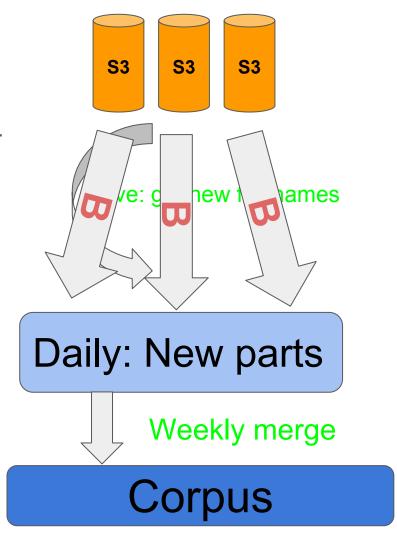
### Use case: Content based user recommendation.

Trained model (single machine): process >1M English stories (use implementation in **Gensim**), corresponding to ~450k users ⇒ vectorize stories & users.

For a given story (distributed): Vectorize its text. Run KNN

## Text import from S3

- Corpus is stored in folders in an S3 bucket.
- Process text (English, Spanish, Tagalog).
- Daily throughput: ~450-500k story chapters (~3GB).
- Outline of the process:
  - Retrieve list of new parts added yesterday.
  - Boto+Spark: Retrieve raw text; process & clean-up text.



# Text import from S3 (continued)

- Previously implemented in Pig + Bash.
- Running-time improvement:
  - Previously: ~3 hours (for English only)
  - Currently: ~1 hour (for three languages; roughly x2 the total amount of text).

### Next steps:

- Add support for all languages.
- Optimize the process by switching to Spark dataframe operations.

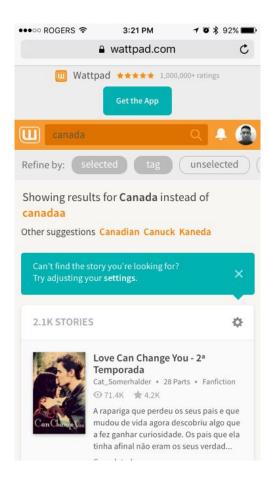
# Search

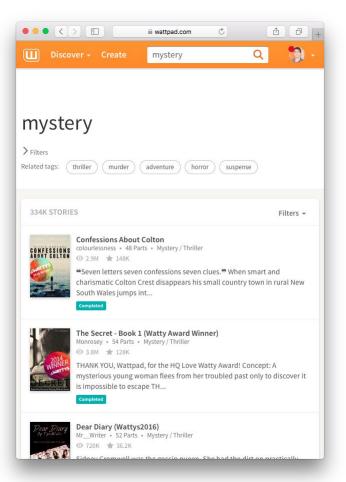
### Search

- Use Elasticsearch to power our search system
- API queries Elasticsearch
- Calculate tf-idf for parts of the documents
  - Titles
  - Descriptions
  - Interaction data: Votes, Reads, etc.
  - Many other fields
- Index is updated as documents change

### Search: Metrics

- Track search and click events on results
- Allows us to calculate performance metrics:
  - o DCG, P@N, etc.
- Use this to identify poor search performance for certain queries
- Use this to run A/B tests on search
- Side note: Use Spark to quickly iterate on these metrics





### Search: Reformulations

- Suggest new query or automatically replace query for common typos
- Sessionize queries
  - Amount of time between actions
  - Dwell time on document
  - String similarity between queries
  - How likely is this query going to result in a click
- Result is a mapping of commonly misspelled queries to correct queries

# Search: Query Suggestions

- Help people browse search by suggesting related queries
- Again using sessionization similar to before
- Vectorize queries and items clicked on using word2vec
- Other searchers where a similar group of stories were clicked on is a good signal

### Search: Reorder results

- Develop a mapping of documents => queries that people use to get to this document
- Sessionization comes in handy here again because we can get more context (other queries in this session)
- This addition signal can be plugged into Elasticsearch as another field and used at query time
- All these jobs are done in batch daily

# Questions?

We're Hiring Come see us!