



Implementation of an exploratory workbench for identifying similar design decisions

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Agenda

- □ Introduction: Comparing Two Decisions
- □ Introduction: Why Compare ?
- Motivation
- Research Questions
- Approach: K-Means
- Observations
- Further Research
- End User System
- □ Configurable Backend System (Pipelines)
- Evaluation Strategy
- Timeline



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In software engineering and software architecture design, architectural decisions (ADs) are design decisions that address architecturally significant requirements; they are perceived as hard to make and/or costly to change.

- Grady Booch, Architecting the unknown, Saturn 2016





Issues	<u>SPARK-8321</u>	<u>SPARK-19625</u>
Description	Authorization Support(on all operations not only DDL) in Spark Sql	Authorization Support(on all operations not only DDL) in Spark Sql version 2.1.0
Concepts	Apache, SQL, authentication	Apache, SQL, authentication
Keywords	Spark, operations, Support, Authorization	Spark, operations, Support, Authorization
Components	SQL	Spark Core, SQL
Issue Type	Improvement	Improvement
Created	12/Jun/15 03:34	16/Feb/17 09:36
Resolved	16/Jun/16 08:22	24/Mar/17 01:21



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Helpful if the second reporter could have been informed about the similar design decision made in past

- Reduced time for analysis
- □ Reduced time to resolution
- Reduced time to turn around for expert feedback

Given an new open design decision, search the knowledge base for similar earlier made design decisions.



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Motivation

- Documentation - specifying constraints on similar design decisions
- Communication visual representation of related design decisions
- Complexity Inferring the complexity for addressing similar design decisions

Research Questions

- How to identify similar design decisions?
- What are the context parameters that needs to be considered?
- Which similarity measures are most efficient for comparing context parameters?



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Approach



Goal: Analyse alternatives for performing text similarity

Machine learning model for unsupervised clustering of design decisions

- □ Predicting cluster label for a new design decision
- □ Ranking within cluster to find most similar design decisions using context



Observations K-Means (Spark and Hadoop Datasets)

With lower K value (k <= 4 clusters) and no pre-processing

- Inconsistent cluster
- □ Large first cluster
- Clustering based on missing values

Lessons learnt → Need pre-processing

With higher K value (k = 8 & k = 20) and with pre-processing

- Uniform clusters
- □ a more uniform spread for 20 clusters
- □ Best Assumption: 8 < K < 20
- □ However, some cluster have <= 7 members,
 - → include member from other cluster in the results ?
 - ➔ Fuzzy C-Means



Further Research

- □ K-Means vs Fuzzy C-Means → Mutually exclusive clusters vs clusters with membership weights
- □ Finding optimum k value
- □ Ranking within clusters → Compare using context similarity measure



End User System Extending AMELIE to Include Similar Design Decisions



Projects Editor Recommender About

Similar/Related Decisions Refactor JDBCRDD to expose JDBC SparkSQL conversion functionality 19 It would be useful if more of JDBCRDD s JDBC Spark_SQL functionality was usable from outside of JDBCRDD this would make it easier to write test harnesses comparing Spark output against other JDBC databases з 17 18 2 root set 13 11 C ANNOTATE Upload area Drop some files here, or click and select files to upload

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Amelie



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Backend System Configurable Pipelines

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Document Clusterer	Create Pipeline	Cluster Docume
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peline: K-r	means		Link SocioCortex workspace		
Select a Workspace	melie 💙				
Select attributes for	mining				
belongs_to	concepts	Description	design decision	patterns	qualityAttributes
Summary	decisionCategory	🗆 Issue Type	Linked Issues	☐ fix version	🗔 status
resolution	🗌 title	decision level	🗌 type	priority	component
🗆 assignee	reporter	created	resolved	🗌 updated	🗌 github_link
☐ resolved pull request	□ source_url				
Select pre-processi	ng steps				
Tokenize	Replace Missing	Filter StopWords	Word2Vec	C Vectorize	ToLowercase
					Update Confi





Qualitative Strategy

□ Expert Evaluation by Employees of Siemens (Experiment Dataset provided by Siemens)

Quantitative Strategy

□ Creating a Test Dataset from Open Source Projects that contains duplicates

□ Evaluate the trained model for precision and recall



Thesis Timeline







Thank you

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End User System: AMELIE The Project Explorer

Projects Editor Recommender Abo

Search ..

Amelie

Projects

Project Name	Description	Category	# Issues
Spark	Apache Spark is a fast and general cluster computi	Apache Software Foundation	899
Hadoop Common	Hadoop Common is the common library for Apache Had	Hadoop	572
Commons CSV		Commons	2
Maven Doxia	Doxia is a content generation framework which aims	Maven	Import
Cocoon	Apache Cocoon is a web development framework built	Cocoon	Import
Triplesec	Triplesec Strong Identity Server: Combined strong	Directory	Import
ServiceMix	ServiceMix is an open source Apache licensed Enter	ServiceMix	Import
Maven Javadoc Plugin	Maven Javadoc Plugin	Maven	Import
CXF-Fediz	Web Application SSO based on WS-Federation	CXF	Import
Kafka	Apache Kafka is a distributed streaming platform.	Kalka	Import
Qpid Proton		Qpid	Import
James Postage		James	Import
Commons Imaging	Renamed from SANSELAN	Commons	Import
TomEE	All-Apache Java EE 6 Web Profile stack based on To	OpenEJB	Import
TinkerPop	TinkerPop: A Graph Computing Framework		Import

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End User System: AMELIE A Visual Frontend

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Observations Detailed K-Means & Similarity

With lower K value (k <= 4 clusters) and no pre-processing

- □ If attribute "design decision" is included, cluster members are those with values 1 & 0 value for it.
- □ If quality attribute is included, cluster members are based on type of quality,
 - Not required, we already have classification based on this.
- □ Clustering based on the summary and description attributes of issues
 - □ Leads to inconsistent clusters with the initial assignment of a one document to each cluster and followed by the assignment of all documents to the first cluster.
- Clustering based on missing values

With higher K value (k = 8 & k = 20) and with pre-processing

- □ Uniform clusters (a more uniform spread for 20 clusters)
- ❑ However, some cluster have <= 7 members, include member from other cluster in the result → Fuzzy C-Means</p>

With Direct Similarity Measure

Equidistant from eachother





Identifying Similar Design Decisions, Prateek Bagrecha (© Florian Matthes, 2017)





1) Set K – To choose a number of desired clusters, K.

2) Initialization – To choose k starting points which are used as initial estimates of the cluster centroids. They are taken as the initial starting values.

3) Classification – To examine each point in the dataset and assign it to the cluster whose centroid is nearest to it.

4) Centroid calculation – When each point in the data set is assigned to a cluster, it is needed to recalculate the new k centroids.

5) Convergence criteria – The steps of (iii) and (iv) require to be repeated until no point changes its cluster assignment or until the centroids no longer move.

