MacroBase: Efficient Explanation On Big Data

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Explaining your data

Problem:

My users are complaining my app crashes a lot.

Explaining your data

Better question:
Which factors in my logs seem related to the crashes?

Explaining your Data

INCHANG Covered Recipient Physician	705877 DANIEL	DORTON	3401 W GORE BLVD	#LAWTON OK	73505-633 United States	Medical D Allopathic OK		Cianna Me	1E+11 Cianna Me CA	Ur
INCHANG Covered Recipient Physician	182729 CAROLINED	SOBOTA	1055 N CURTIS RD	BOISE ID	83706-135 United States	Medical D Allopathic ID		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	875992 RODOVALDO	RODRIGUEZ	1504 N TH SUITE 101	DALTON GA	30720-839 United States	Medical D Allopathic GA		Cianna Me	1E+11 Cianna McCA	Ur
INCHANG Covered Recipient Physician	420837 KATHRYN S	NORTON	1100 N 19 SUITE 4G	ABILENE TX	79601-234 United States	Medical D Allopathic LA	TX	Cianna Me	1E+11 Cianna McCA	Ur
INCHANG Covered Recipient Physician	318239 MICHAEL J	BURRELL	2525 W UNSUITE 403	MUNCIE IN	47303-340 United States	Medical D Allopathic IN		Cianna M€	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	136708 JONATHAN	TAY	6630 B S N RAD ONC	RENO NV	89509 United States	Medical D Allopathic NV		Cianna Me	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	45955 MARK S	СНОН	3245 GRO\#202	BERWYN IL	60402-347 United States	Medical D Allopathic IL		Cianna Me	1E+11 Cianna M∈CA	Ui
INCHANG Covered Recipient Physician	1354132 ELLIOT BRIN	NAVO	1121 KINNEYS LN	PORTSMO OH	45662-280 United States	Medical D Allopathic OH		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	183467 KORY	JONES	811 W INT STE 212	ARLINGTOTX	76017-587 United States	Medical D Allopathic TX		Cianna Me	1E+11 Cianna M∈CA	Ui
INCHANG Covered Recipient Physician	46859 DAVID H	KIM	3699 EPWORTH RD	NEWBURGIN	47630-890 United States	Medical D Allopathic IN		Clanna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	1694 STEVEN E	OLYEJAR	695 S DOBSON RD	CHANDLEIAZ	85224-566 United States	Medical D Allopathic AZ		Cianna Me	1E+11 Cianna McCA	Ui
INCHANG Covered Recipient Physician	1104947 ANDERSO A	BAUER	18037 N 94TH WAY	SCOTTSDA AZ	85255-608 United States	Medical D Allopathic AZ		Clanna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	1104947 ANDERSO A	BAUER	18037 N 94TH WAY	SCOTTSDA AZ	85255-608 United States	Medical D Allopathic AZ		Cianna Me	1E+11 Cianna McCA	Ui
INCHANC Covered Recipient Physician	1104947 ANDERSO A	BAUER	18037 N 94TH WAY	SCOTTSDA AZ	85255-608 United States	Medical D Allopathic AZ		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	274809 PATRICIA L	CLARK	12701 WE! SUITE 250	HOMER GUL	60491 United States	Medical D Allopathic IL		Cianna Me	1E+11 Cianna MeCA	Ui
INCHANC Covered Recipient Physician	1218858 NOAH	WEMPE	2200 W ILLINOIS AV	EMIDLAND TX	79701-640 United States	Medical D Allopathic TX		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	905501 AMRIT	MANGAT	430 PENN: SUITE 350	GLEN ELLY IL	60137-446 United States	Medical D Allopathic IL		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	92700 MARK FREDRICK	SCHRAY	1015 NW 2LL50	PORTLANIOR	97210-302 United States	Medical D Allopathic OR		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	319661 ALVARO H	DEVIA	6554 S MC #B	RENO NV	89509-611 United States	Medical D Allopathic NV		Cianna Me	1E+11 Cianna MeCA	U
INCHANC Covered Recipient Physician	364961 CHERYL	STANSKI	105 W STC STE 4A	KINGSPOFTN	37660-325 United States	Medical D Allopathic TN		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	1296080 MICHAEL	HAILEY	9000 AIRLI STE #250	BATON RCLA	70815-411 United States	Medical D Other Ser LA		Cianna Me	1E+11 Cianna MeCA	U
INCHANG Covered Recipient Physician	93848 WALTON ALBERT	TAYLOR	7777 FOREST LN STE	DALLAS TX	75230-683 United States	Medical D Allopathic TX		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	93848 WALTON ALBERT	TAYLOR	7777 FOREST LN STE	DALLAS TX	75230-683 United States	Medical D Allopathic TX		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	139417 HANK C	HILL	4205 BELF SUITE 200	SJACKSON\FL	32216-587 United States	Medical D Allopathic FL		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	678599 CARLOS	LOPEZ	624 S TONOPAH DR	LAS VEGA! NV	89106-402 United States	Medical D Allopathic FL	NV	Cianna Me	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	763932 ERICA M.	GIBLIN	2000 MED SUITE 200	ANNAPOL MD	21401-374 United States	Medical D Student, FMD		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	792203 FAISAL AMIN	SIDDIQUI	400 NE MOTHER JOS	EVANCOUV WA	98664-320 United States	Medical D Allopathic OR	WA	Cianna Me	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	735617 NAMRATA	SETHI	401 W POI PROVIDE	WALLA W. WA	99362-284 United States	Medical D Allopathic WA		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	479406 JOANNE MARIE	TRAPENI	35 MONUI STE 201	YORK PA	17403-502 United States	Medical D Allopathic PA		Cianna Me	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	621360 ABHINANI VENKATA	PEDDADA	1155 MILL ST	RENO NV	89502-157 United States	Medical D Allopathic CA	NV	Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	1335579 GUY CLEAGE	JONES	101 W 8TH AVE	SPOKANE WA	99204-230 United States	Medical D Allopathic WA		Cianna Me	1E+11 Cianna M∈CA	U
INCHANG Covered Recipient Physician	321647 NEWTON CRAIG	BRACKETT	4181 HWY COASTAL	MURRELLSSC	29576 United States	Medical D Allopathic SC		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	321647 NEWTON CRAIG	BRACKETT	4181 HWY COASTAL	MURRELLSSC	29576 United States	Medical D Allopathic SC		Cianna Me	1E+11 Cianna M∈CA	U
NCHANG Covered Recipient Physician	230505 ALEXANDITUAN	ROSE	1887 KING STE 1900	ORANGE F FL	32073-441 United States	Medical D Student, FFL		Cianna Me	1E+11 Cianna McCA	U
NCHANC Covered Recipient Physician	275722 WILLIAM H	TURNEY	6600 FISH SUITE 101	WACO TX	76710-258 United States	Medical D Allopathic TX		Cianna Me	1E+11 Cianna McCA	U
INCHANG Covered Recipient Physician	679547 SIGMUND ALLEN	AMITIN	1203 WEST SUITE C	ANNAPOL MD	21401-366 United States	Medical D Allopathic MD		Cianna Me	1E+11 Cianna McCA	U
JNCHANC Covered Recipient Physician	423113 MELODY	PAULISHAK	500 UNIVERSITY DR	HERSHEY PA	17033-236 United States	Medical D Allopathic PA		Cianna Me	1E+11 Cianna McCA	U
NCHANC Covered Recipient Physician	322013 JOHN P	WILLIAMS	8650 SUDL #206	MANASSA VA	20110 United States	Medical D Allopathic VA		Cianna Me	1E+11 Cianna McCA	U
INCHANC Covered Recipient Physician	186844 ANNE	O'NEILL	8930 W SU SUITE#30	LAS VEGA! NV	89148-500 United States	Medical D Allopathic NJ		Cianna Me	1E+11 Cianna M∈CA	U

I don't know what's going on here, and neither do you.

Explaining your Data

Version	Country	Support	Ratio
4.1	France	0.50	4.0

Here, Version 4.1 in French is the biggest difference between crashing and successful logs.

Explanations

Attribute 1	Attribute 2	Attribute 3	Metric 1	Metric 2
Value 1	Value 2	Value 3	X.XX	Y.YY
•••	•••		•••	

MacroBase returns explanations, sets of attributes that explain difference between two datasets.

What is MacroBase?

- MacroBase introduces a new SQL operator, DIFF.
- DIFF helps you find differences between data.
- DIFF is implemented using Spark and Spark-SQL.

TableSuccess

TableCrash

Version	Carrier	Country
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier	Country
4.1	Verizon	France
4.0	Verizon	USA

Here are some logs. Let's figure out what's causing the crashes

TableSuccess

TableCrash

Version	Carrier	Country
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier	Country
4.1	Verizon	France
4.0	Verizon	USA







Version	Country	Support	Ratio
4.1	France	0.5	4.0

We want an *explanation*—some attributes that are correlated with the crashes and metrics to tell us how.

TableSuccess

TableCrash

Version	Carrier	Country
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier	Country
4.1	Verizon	France
4.0	Verizon	USA

DIFF

Let's make a new SQL operator that finds explanations!

TableSuccess

TableCrash

Version	Carrier	Country
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier	Country
4.1	Verizon	France
4.0	Verizon	USA

DIFF TableCrash, TableSuccess ON Version, Carrier, Country

DIFF compares the crashing and successful logs. But how does it know how?

TableSuccess

TableCrash

Version	Carrier Country	
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier Country	
4.1	Verizon	France
4.0	Verizon	USA

DIFF TableCrash, TableSuccess
ON Version, Carrier, Country
COMPARE BY

We need to give DIFF a list of rules.

TableSuccess

TableCrash

Version	Carrier	Country	
4.0	Verizon	USA	
4.0	Verizon	USA	

Version	rsion Carrier Country	
4.1	Verizon	France
4.0	Verizon	USA

We want attributes that occur much more often in crashing runs than in non-crashing runs

TableSuccess

TableCrash

Version	Carrier Country	
4.0	Verizon	USA
4.0	Verizon	USA

Version	Carrier Country	
4.1	Verizon	France
4.0	Verizon	USA

We also want those attributes to have high sample sizes, or we'll just get low-sample size noise

TableSuccess

TableCrash

Version	Carrier	Country	
4.0	Verizon	USA	
4.0	Verizon	USA	

Version	Carrier Country	
4.1	Verizon	France
4.0	Verizon	USA

Version	Country	Support	Ratio
4.1	France	0.50	4.0

Anatomy of DIFF

Two tables to DIFF over. Must share same schema.





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Two tables to DIFF over. Must share same schema.







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Two tables to DIFF over. Must share same schema.







Attributes
must
appear in
both
input
tables.

We provide pre-defined difference metrics to use in DIFF. It's also easy to define your own.

Demo - Startup

```
kraftp@pk-macrobase-spark-demo-m: ~/macrobase
                                                                                               \times
{HTTP/1.1,[http/1.1]}{0.0.0.0:4040}
kraftp@pk-macrobase-spark-demo-m:~/macrobase$ spark-submit --master yarn --deploy-mode client -
driver-memory 5g --executor-cores 1 --num-executors 5 --executor-memory 4g --class edu.stanfor-
d.futuredata.macrobase.sql.MacroBaseSQLRepl sql/target/macrobase-sql-1.0-SNAPSHOT.jar -d -n 5
18/06/03 15:51:09 INFO org.spark project.jetty.util.log: Logging initialized @3054ms
18/06/03 15:51:09 INFO org.spark_project.jetty.server.Server: jetty-9.3.z-SNAPSHOT
18/06/03 15:51:09 INFO org.spark project.jetty.server.Server: Started @3140ms
18/06/03 15:51:09 INFO org.spark project.jetty.server.AbstractConnector: Started ServerConnecto
r@54534abf{HTTP/1.1,[http/1.1]}{0.0.0.0:4040}
18/06/03 15:51:09 INFO com.google.cloud.hadoop.fs.gcs.GoogleHadoopFileSystemBase: GHFS version:
1.6.5-hadoop2
18/06/03 15:51:10 INFO org.apache.hadoop.yarn.client.RMProxy: Connecting to ResourceManager at
pk-macrobase-spark-demo-m/10.138.0.33:8032
18/06/03 15:51:12 INFO org.apache.hadoop.yarn.client.api.impl.YarnClientImpl: Submitted applica
tion application 1528040625320 0006
Welcome to MacroBase!
macrobase-sql>
```

Demo - Ingest

kraftp@pk-macrobase-s	park-demo-m: ~/	macrobase	- □ ×	
macrobase-sql> IMPORT FROM CSV FILE 'PGYR-small.csv' INTO cms(Recipient_State string, Applicable_M ^ anufacturer_or_Applicable_GPO_Making_Payment_Name string, Name_of_Associated_Covered_Drug_or_Biolo gical1 string, Program_Year string); 18/06/03 16:06:24 INFO edu.stanford.futuredata.macrobase.sql.macroBaseSQLKepl: ImportCsv{filename= PGYR-small.csv, tableName=cms, columns={Applicable_Manufacturer_or_Applicable_GPO_Making_Payment_N ame=STRING, Program_Year=STRING, Name_of_Associated_Covered_Drug_or_Biological1=STRING, Recipient_ State=STRING}}				
18/06/03 16:06:24 IN	NFO org.apac	ne.hadoop.mapre	ed.FileInputFormat: Total input files to process : 1	
	olicable_Man	ufacturer_or_Ap	plicable_GPO_Making_Payment_Name Name_of_Associated_	
+				
TX		2042	Zimmer Holding Inc	
TX	null	2013	Zimmer Holding Inc	
OR	null	2013	LifeCell Corporation	
	null	2013		
sc	null	2013	Edwards Lifescien	
CA			Edwards Lifescien	
IL	null	2013	Zimmer Holding Inc	
	null	2013		
MA	Rituxan	2013	Genentech USA, Inc.	
TX	Rituxan	2013	Genentech USA, Inc.	

Demo – DIFF

```
kraftp@pk-macrobase-spark-demo-m: ~/macrobase
macrobase-sql> SELECT * FROM DIFF (SPLIT cms WHERE Program Year="2015") ON Recipient State, Applic
able Manufacturer or Applicable_GPO_Making_Payment_Name, Name_of_Associated_Covered_Drug_or_Biolog
ical1 WITH MIN RATIO 1.3 MIN SUPPORT 0.02 COMPARE BY globalratio(COUNT(*));
18/06/03 16:10:41 INFO edu.stantord.tuturedata.macrobase.sql.MacroBaseSQLRepl: Query{queryBody=Dit
fQuerySpecification{select=Select{distinct=false, selectItems=[*]}, first=Optional.empty, second=n
ull, attributeCols=[Recipient_State, Applicable Manufacturer or Applicable GPO Making Payment Name
, Name of Associated Covered Drug or Biological1], minRatioExpr=DECIMAL '1.3', minSupportExpr=DECI
MAL '0.02', ratioMetricExpr=globalratio(COUNT(*)), maxCombo=3, where=null, orderBy=Optional.empty,
limit=null, exportExpr=null}}
18/06/03 16:10:53 INFO AttributeEncoderDistributed: Column cardinalities: [17, 13, 4]
18/06/03 16:11:04 INFO APLSummarizerDistributed: Encoded in: 23572
18/06/03 16:11:04 INFO APLSummarizerDistributed: Encoded Categories: 34
18/06/03 16:11:05 INFO APLSummarizerDistributed: Time spent in order 1:
                                                                        443
18/06/03 16:11:05 INFO APLSummarizerDistributed: Time spent in order 2:
                                                                        197
18/06/03 16:11:05 INFO APLSummarizerDistributed: Time spent in order 3:
                                                                        150
18/06/03 16:11:05 INFO APLSummarizerDistributed: Number of results: 2
Applicable Manufacturer or Applicable GPO Making Payment Name Recipient State Name of Associated
                                       support| global ratio|outlier count|total count|
Covered Drug or Biological1
                                           Actavis Pharma Incl
                                                                         null
                      null | 0.03360490895238618 | 1.323027911511267 |
                                                                         4896.0
                                                                                     5080.0
                                               Lilly USA, LLC
                                                                         null
                      null|0.026109696416437303|1.372749548708586|
                                                                         3804.0
                                                                                     3804.0
```

How DIFF works: High Level

- DIFF calculates difference between two tables
- Quantify difference with difference metrics
- Return all sets of attributes that pass difference metrics

First, iterate through singleton attributes and store their frequency counts (or other aggregates)

Version	Carrier	Country	Crash
4.0	Verizon	USA	False
4.1	Verizon	USA	False
4.0	Verizon	France	True

Order 1: {(4.0): (1, 1), (France): (1, 0)...}

Crashing Frequency Count

Success Frequency Count

Next, compute difference metrics—here, support and risk ratio—from aggregates. Assume 1000 logs, 100 of which crashed.

```
Order 1: {(4.0):(9, 691), (France):(70, 330)...}
```

```
Order 1: {(4.0):(0.09, 0.05), (France):(0.7, 2.12)...}
```

Support

Risk Ratio

Then, prune with support (0.1) and risk ratio (3.0) thresholds. Keep track of who passed support threshold even if they failed risk ratio.

```
Order 1: {(4.0):(0.09, 0.05), (France):(0.7, 2.12)...}
```

```
Order 1: {<del>(4.0):(0.09, 0.05</del>),
<del>(France):(0.7, 2.12)</del>...}
```

Now repeat for pairs. Iterate through pairs of attributes where both passed support threshold, even if they failed risk ratio.

Version	Carrier	Country	Crash
4.0	Verizon	USA	False
4.1	Verizon	USA	False
4.1	Verizon	France	True

```
Order 2: {(Verizon, France): (1, 1), (4.1, France): (1, 0)...}
```

Crash Frequency Count

Success Frequency Count

Repeat the same steps and we'll get our answer from before!

Order 2: {(4.1, France):(0.5, 4.0)...}

Version	Country	Support	Ratio
4.1	France	0.50	4.0

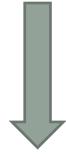
Distribute With Spark-Partitioning

Version	Carrier	Country	Crash
4.0	Verizon	USA	False
4.1	Verizon	France	False
4.0	Sprint	USA	False





Version	Carrier	Country	Crash
4.0	Verizon	USA	False



Version	Carrier	Country	Crash
4.0	Sprint	USA	False

Version	Carrier	Country	Crash
4.1	Verizon	France	False

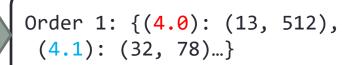
First, partition your data by row to different machines.

Distribute With Spark-Mapping

Version	Carrier	Country	Crash
4.0	Verizon	USA	False

```
Order 1: {(4.0):(11, 522), (4.1): (50, 80)...}
```

Version	Carrier	Country	Crash
4.1	Verizon	France	True



Version	Carrier	Country	Crash
4.0	Sprint	USA	False

```
Order 1: {(4.0): (20, 566), (4.1): (45, 68)...}
```

Then, at each order, make a generalized Apriori map per partition. This is Spark's map step.

Distribute With Spark-Reducing

```
Order 1:{(4.0):(11, 522),

(4.1): (50, 80)...}

Order 1:{(4.0):(13, 512),

(4.1):(32, 78)...}

Order 1:{(4.0):(44, 1600),

(4.1):(127, 226)...}

Order 1:{(4.0):(20, 566),

(4.1):(45, 68)...}
```

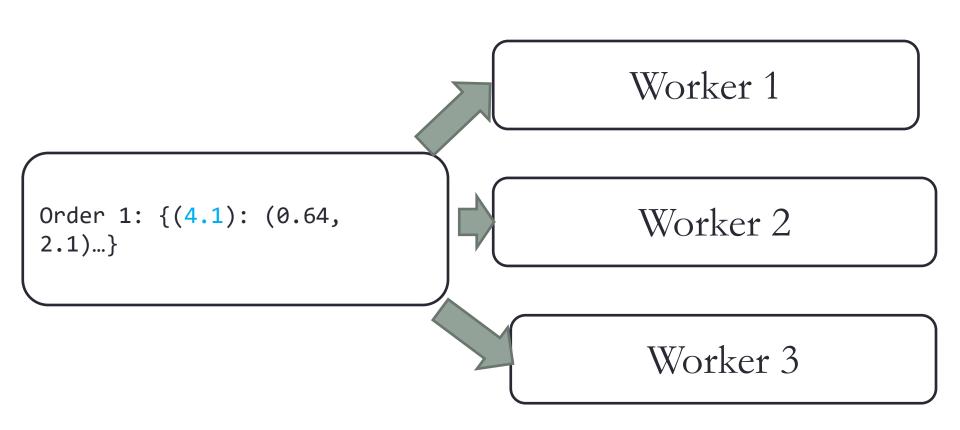
Next, combine all the maps onto a single machine. This is Spark's reduce step.

Distribute With Spark-Finishing

```
Order 1:{(4.0):(44, 1600), (4.1):(128, 226)...}
Order 1: {(4.0): (0.22, 0.4), (4.1): (0.64, 2.1)...}
Order 1: { (1.0). (0.22, 0.1), (4.1): (0.64, 2.1)...}
```

Now that the map is on a single machine, compute difference metrics and prune as normal.

Distribute With Spark-Reducing



Next, combine all the maps onto a single machine. This is Spark's reduce step.

Use Macrobase!

- All our code is open-sourced on GitHub:
- https://github.com/stanford-futuredata/macrobase
- We also have tutorials and guides up on our website:
- https://macrobase.stanford.edu/
- https://macrobase.stanford.edu/docs/sql/spark/
- This includes info for both single-node MacroBase and MacroBase-Spark
- As long as you can get your data into a CSV (and onto HDFS for MB-Spark), you can run MacroBase on it!