KNIME & Avira, or how I've learned to love Big Data

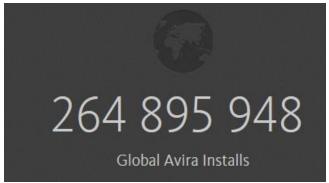






Facts about Avira (AntiVir)

- 100 mio. customers
- "Extreme Reliability"



13.02.2014, 06:58

 500 employees (Tettnang, San Francisco, Kuala Lumpur, Bucharest, Amsterdam)

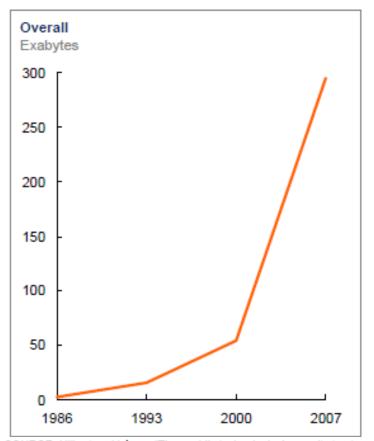
Company owner, Mr. Auerbach

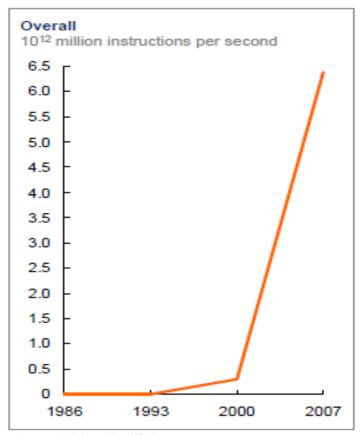


Big Data: Why did Avira decided to invest?

Data storage has grown significantly after 2000

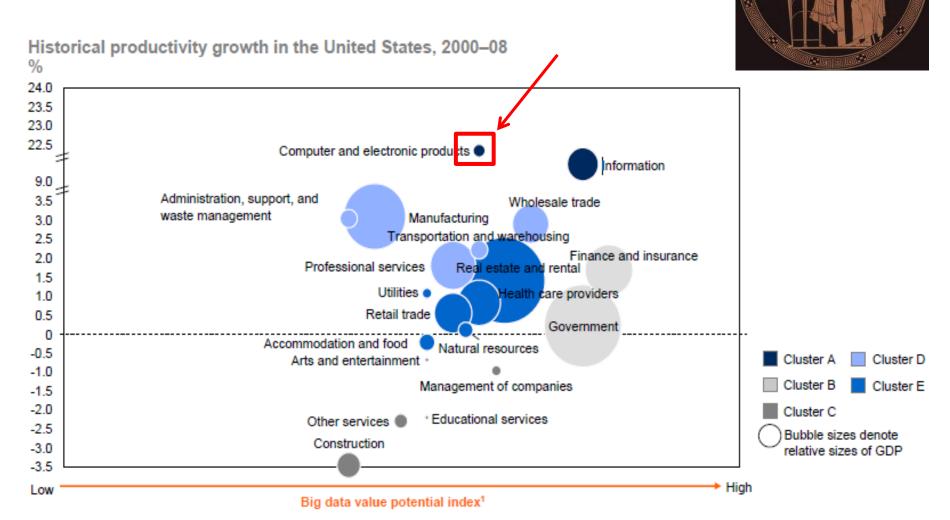
Computation capacity has also risen sharply





SOURCE: Hilbert and López, "The world's technological capacity to store, communicate, and compute information," Science, 2011

Reason 1: Because the McKinsey prophets said it may bring us more money ©



SOURCE: US Bureau of Labor Statistics; McKinsey Global Institute analysis

Reason 2: Because we are sitting on the top of several dozens of terabytes of anonymized customer data.

The only challenge was to link this data and converge towards a single customer profile

Documented Datasources, unique ID & data volume		
	Protection Cloud uID = rndsnr; Volume = 10gb/day	
LogBox (Product	: Download & VDF Update): uID = rndsnr & license no; Volume = 30 gb/day	
Global	Mailing System: uID = license no.; Volume = 7,5 mio Emails/month	
	eshop.avira.com: uID = license no.; Volume = 10 gb/day	
Licensing	system: uID = license no.; Volume = registration data100 mio users	
Clev	erbridge Shopping Cart: uID = license no.; Volume = 10gb/day	
Website Site (Catalyst: Uid not yet defined; Volume = 65 mio page loads & clicks/month	
No.	otifier: Uid not yet defined; Volume = 20 mio impressions/day	
	IPM: Uid not yet defined = 12,5 mio impressions/day	

Reason 3: Because we want to perform customer profilling & next best offer marketing to increase our margin

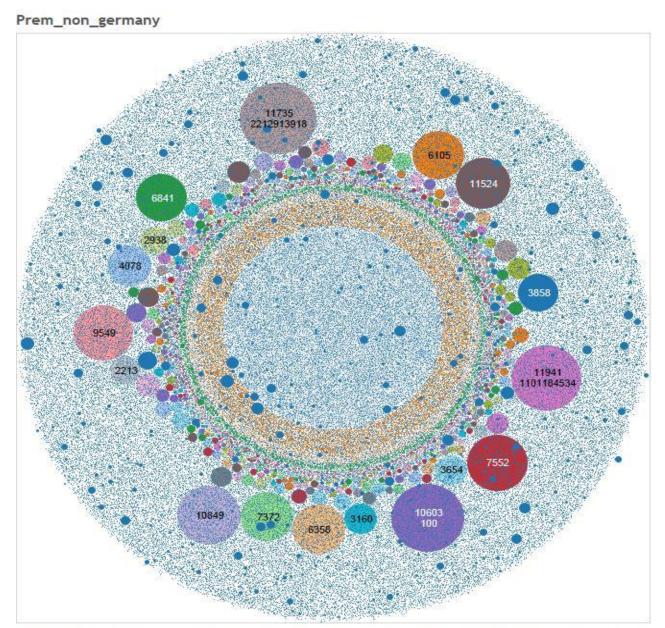
Business objectives

- Deliver the optimal price, features & messaging to each customer in order to maximize EBIT from sales of new and renewal licenses
- Protect long-term margins by making each offer timely and unrepeatable (e.g., unique to a specific customer, product, event)
- Learn the underlying mechanics of features and price-elasticity on the level of customer cells
- Enable business to optimize campaign portfolio (i.e., über-algorithm traffics campaigns)
- A/B testing in all customer touch-points

Technology requirements

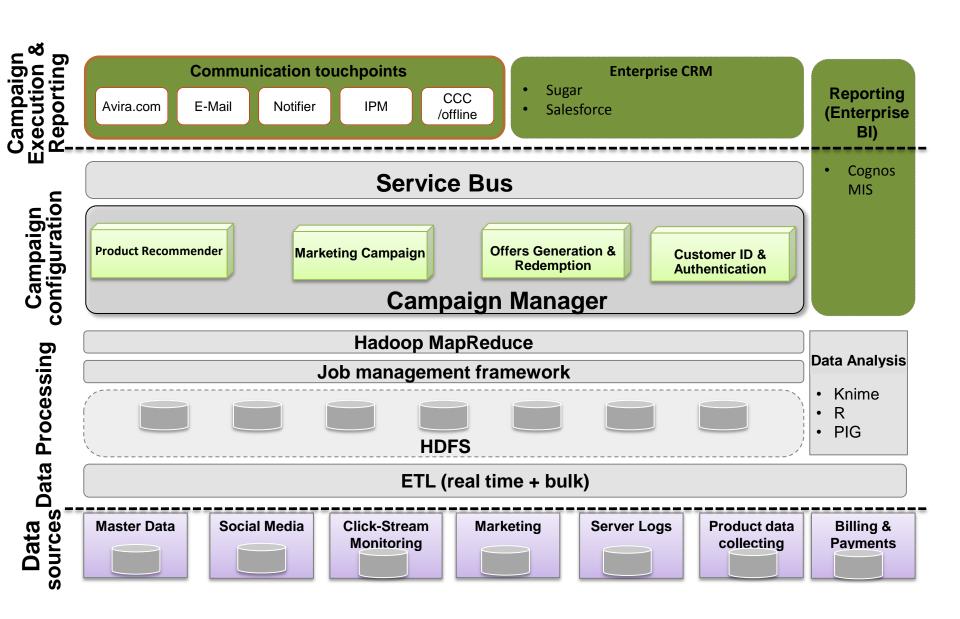
- Ability to link a specific offer in a specific touch-point to specific customer.
- Machine-learning over all design dimensions to continually improve performance of the application
- Causal reporting to maximize the learning effect in the organization from algorithmic approaches to marketing and automation
- Create customer, product and behavioral tables from Avira's raw data within the dev. environment
- Create machine-learning algorithms optimizing the offer (price, features) per user-session
- Implement the services on development platform and place in listening mode to train
- Setup of the Hadoop framework (HDFS & MapReduce) & Couchbase, KNIME & Impala

Reason 4: Because it's fun (isn't this beautiful?)



Counts of rndsnr and license. Color shows details about counts of rndsnr. Size shows sum of total updates . The marks are labeled by counts of rndsnr and license.

Our business/architectural vision



Our Daily Data

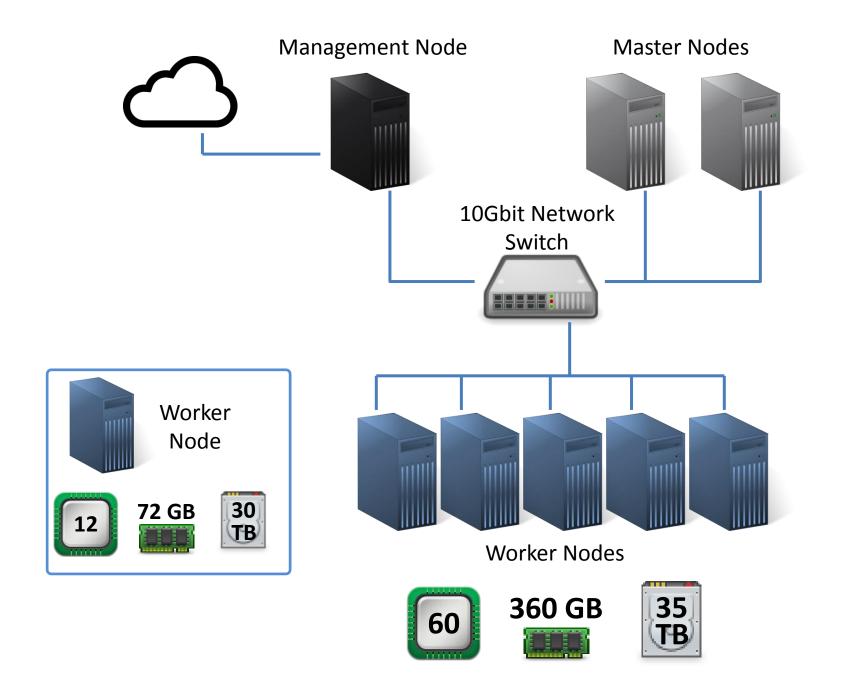
- Website logs: 5.000.000 lines
- Installation logs: 2.200.000 lines
- InProduct Messaging: 43.000.000 lines
- Download/Updater logs: 2.000.000.000 lines

60GB of daily compressed data or around 2TB of monthly data

Finding the Right Tool

- Our data is:
 - Unstructured; Messy
 - Coming from all kinds of sources:
 log files, log tables, relational databases
- We want to:
 - Gather and store historical data
 - Process huge amounts of it
 - Support both batch and real-time operations







CDH

BATCH PROCESSING (MapReduce, Hive, Pig) ANALYTIC SQL (Impala) SEARCH ENGINE (Cloudera Search)

MACHINE LEARNING (Spark, MapReduce, Mahout) STREAM PROCESSING (Spark)

3RD PARTY APPS (Partners)

WORKLOAD MANAGEMENT (YARN)

STORAGE FOR ANY TYPE OF DATA

UNIFIED, ELASTIC, RESILIENT, SECURE (Sentry)

Filesystem (HDFS) Online NoSQL (HBase)

DATA INTEGRATION (Sqoop, Flume, NFS)

Using The Right Tool









MapReduce



- Full control over how the data is processed
- Works on structured and unstructured data
- Good for very complex business logic

- You have to write Java code
- Restricted to the MapReduce programming model
- Some things are difficult to code (JOINs, custom sorting)

Hive



- You write SQL-like queries
- Great for ad-hoc queries, data exploration
- Very fast development

- Works only on structured data
- Gets ugly if the business logic is complex

Impala



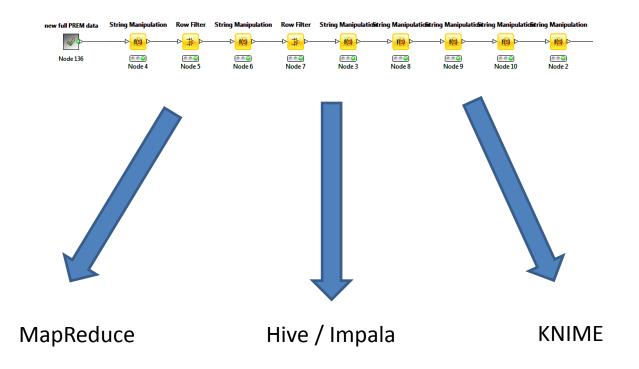
- Like Hive but A LOT faster
- Runs directly in memory
- Delivers almost real-time results

- Limited to in-memory processing
- Unreliable

KNIME and Hadoop at Avira

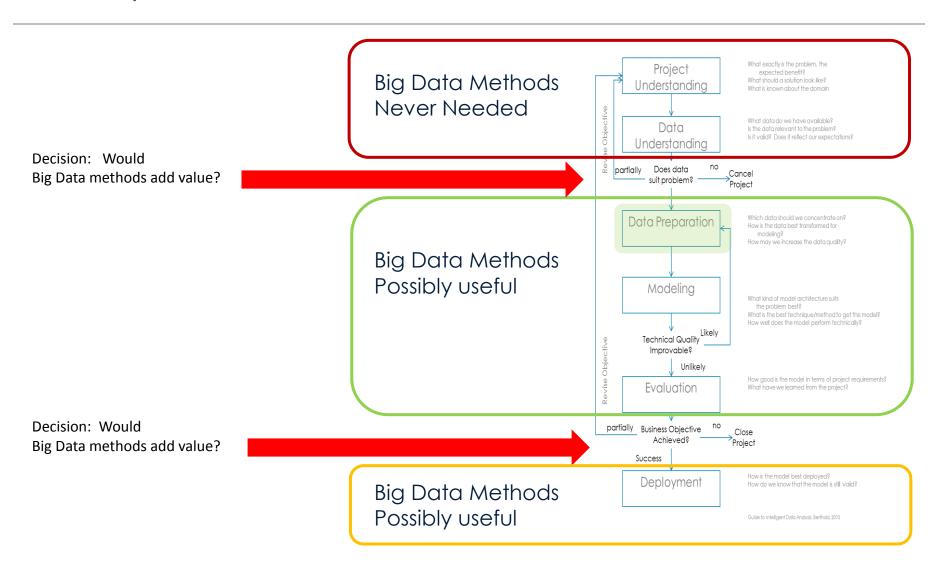
*0: 004003_Database_QueryBuildWithQuickforms 🔀 Jdbc: Hive Table + **Row Count** Userid Impala **SQL Statement** Limit = 100000 Password Database Connection Reader Evaluate / Override SQL Access Node 22 Node 19 Node 21 Node 1

Knime for Understanding the Data



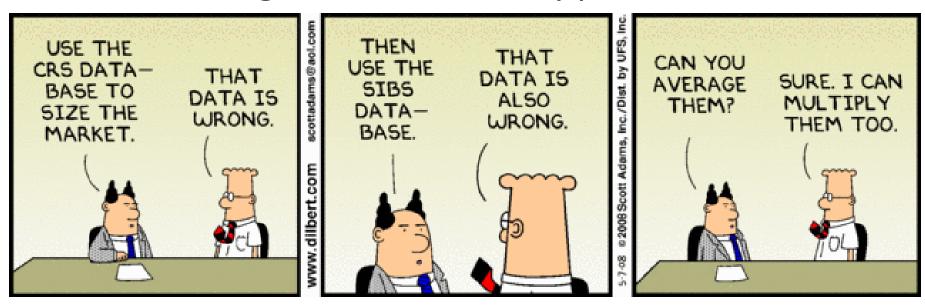
Some Big Data Practicalities

Use Crisp-DM!



Why did we decided to go with KNIME?

 The dark side of the moon: a typical symptom for home-grown business applications.



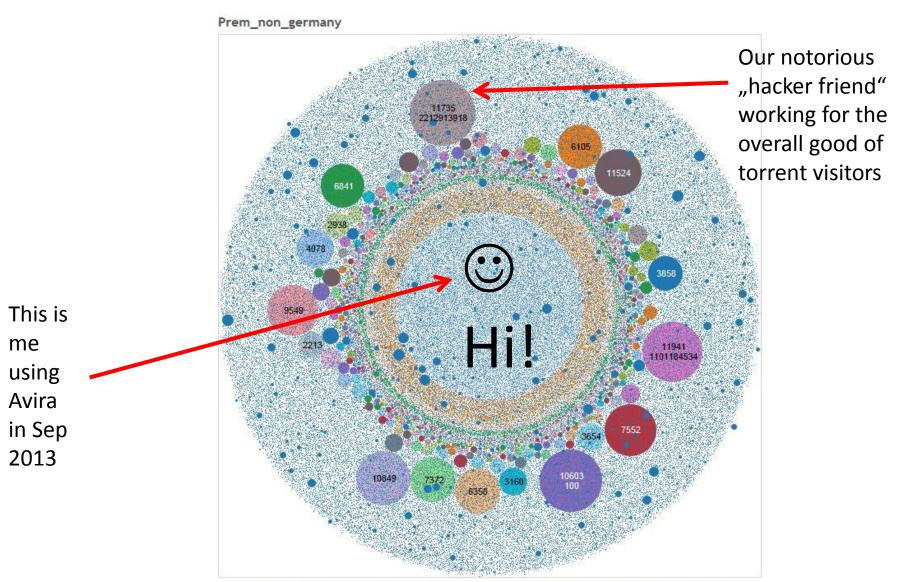
KNIME helped us to tie a "knot" for the multiple uncorrelated data points and create customer 360 tables

Why did we decided to go with KNIME?

It helped us move from code based data mining towards workflow based analytics; Analytics for everyone, easy to explain to all management/company levels

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RCode["function (x, na.last = TRUE, ties.method = c(\"average\", \"first\",
            \"random\", \"max\", \"min\")) \n{\n nas <- is.na(x)\n
                           ties.method <- match.arg(ties.method)\n
     nm <- names(x)\n
     if (is.factor(x)) \n
                                        x <- as.integer(x)\n
     switch(ties.method, average = , min = , max = .Internal(rank(x[!nas],
     \n
                 ties.method)), first = sort.list(sort.list(x[!nas])),
     \n
                 random = sort.list(order(x[!nas],
     stats::runif(sum(!nas)))))n if (!is.na(na.last) && any(nas))
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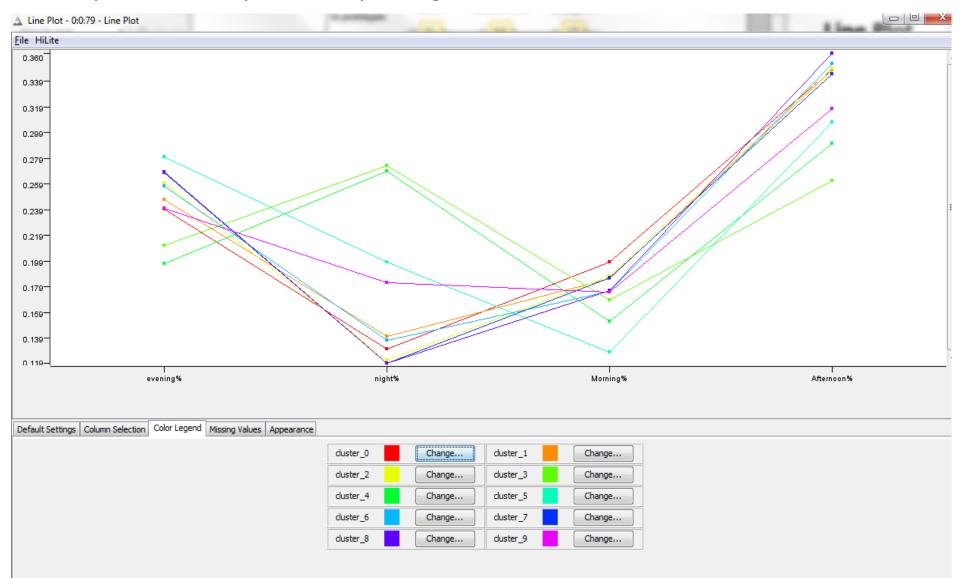
Some results: using KNIME & Tableau we've managed to perform forensycs and license outlier analysis



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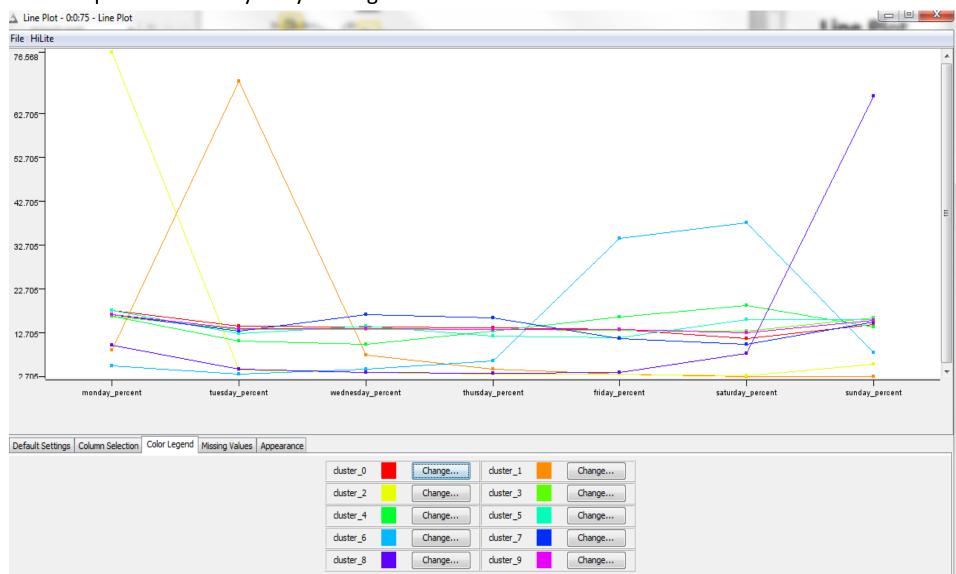
Running k-means in KNIME to identify relevant clusters for Germany by looking at their antivirus software update behaviour

Graph of clusters by time of day % usage:



Running k-means in KNIME to identify relevant clusters for Germany by looking at their antivirus software update behaviour

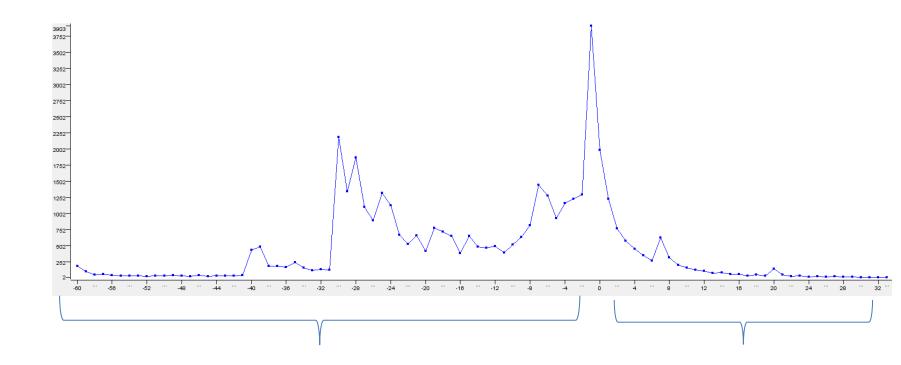
Graph of clusters by daily % usage



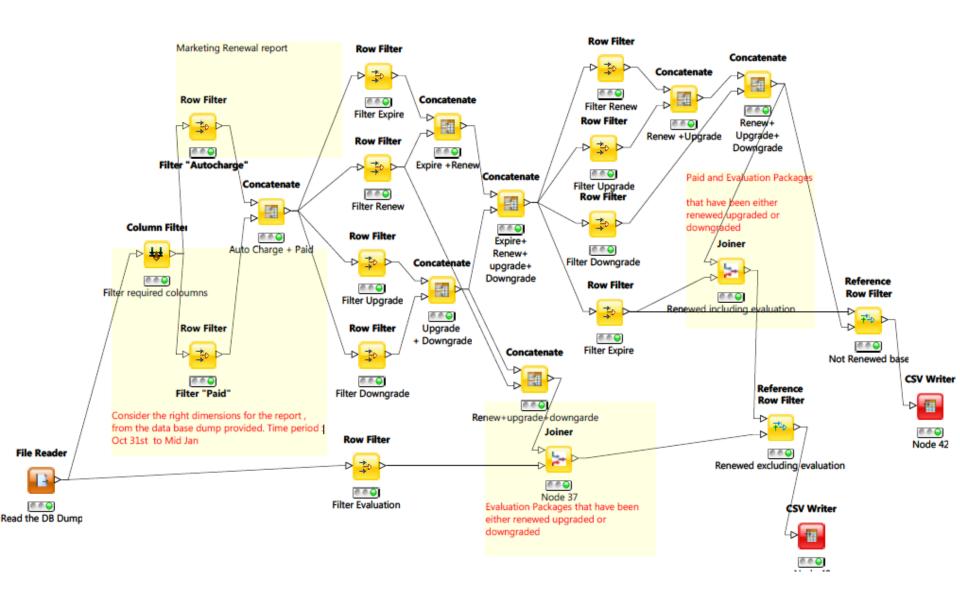
Using KNIME to identify the real License Renewal pattern of our customers

The timing of renewing an Avira license in %

"Reason 6" = 12%

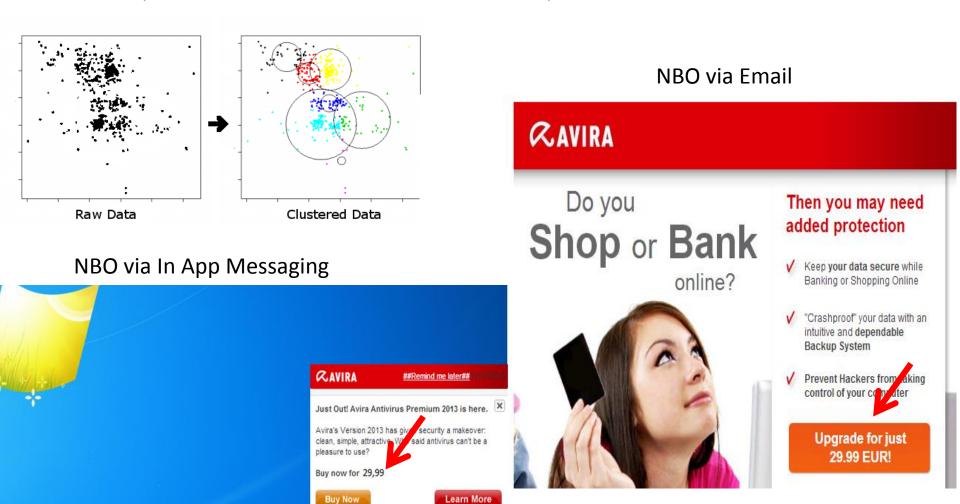


Using KNIME to do standardized reporting of our license renewal metrics



Next steps

 Identify unknown groups of customers by allowing the machine to find patterns in data for creating special association rules/product recommendations & next best offer; test & train in KNIME, real-time model execution in Couchbase;



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Thank you.





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