

Session 11413 Issues in Big Data: Analytics

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Abstract

- **OK, so you've got all the data lying around and you've decided to make it work for you. Now, what you want to know is how to make sense of it. How do you do that? What do you get? Then, what do you do with it once you've gotten it? How do you know if the analysis is to be trusted?**
- **The speakers will address those issues, present IBM best practices, and discuss some real-world use cases.**

Agenda

1

Big Data – Definitions & A Quick Story

2

Technology Primer

3

Dive on Content Analytics

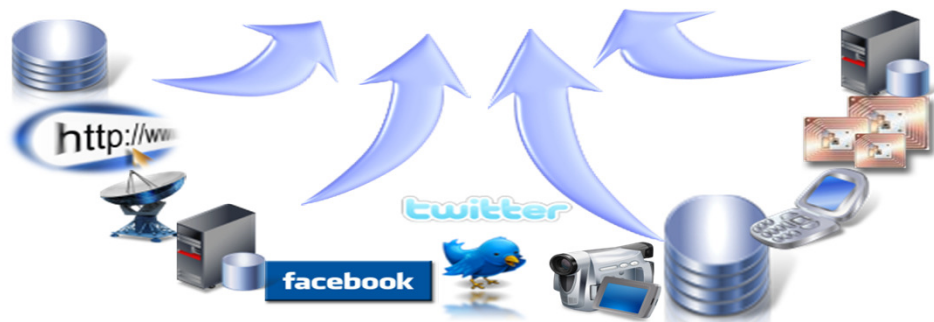
4

Thoughts On Getting Started

Conventional Definition of “Big Data”

- Never before possible
- Solely defined by large volumes
 - Only Unstructured Data
- Valuable insight, but difficult to extract
 - Basically an ETL environment

This definition is wrong



Functional Big Data Definition

Dealing with information management **challenges** that don't natively fit with traditional approaches to handling the problem

Functional Big Data Definition, Part 2

The technologies that deal with these problems are **broad and diverse, it is not just Hadoop**

Giving Rise to “Fit For Purpose Architectures”

The introduction of **purpose-built technology** focused on a **specific computing problem** that is compelling better than existing technologies

Giving Rise to “Fit For Purpose Architectures”

Match the compute problem to the best way to handle it **rather than assuming SQL by default**

...Which In-turn Enables Paradigm Shifts*

- **The Idea Of the Super-set**
- **Move from Sampling to “Absolute Knowledge”**
- **Combining Structured and Unstructured Analytics**
- **Importance of Augmented Decision Making**
- Streaming Computing Paradigm
- Lowering The Cost of Experimentation
- Changing Role Of Archiving

(* but the Laws of Gravity Still Apply!)

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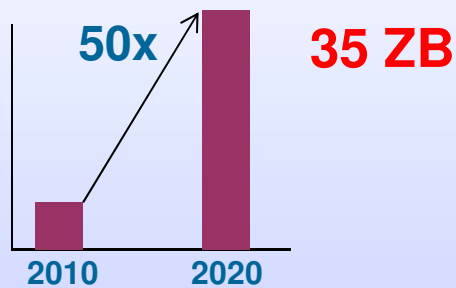
Dive on Content Analytics

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Thoughts On Getting Started

The characteristics of big data

Cost efficiently processing the growing **Volume**



Responding to the increasing **Velocity**



30 Billion RFID sensors and counting

Collectively Analyzing the broadening **Variety**



80% of the worlds data is unstructured



Establishing the **Veracity** of big data sources

1 in 3 business leaders don't trust the information they use to make decisions

Hadoop is Well Suited for Handling Certain Types of Big Data Challenges

Analyzing larger volumes may provide better results



Deriving new insights from combinations of data types



Larger data volumes are cost prohibitive with existing technology



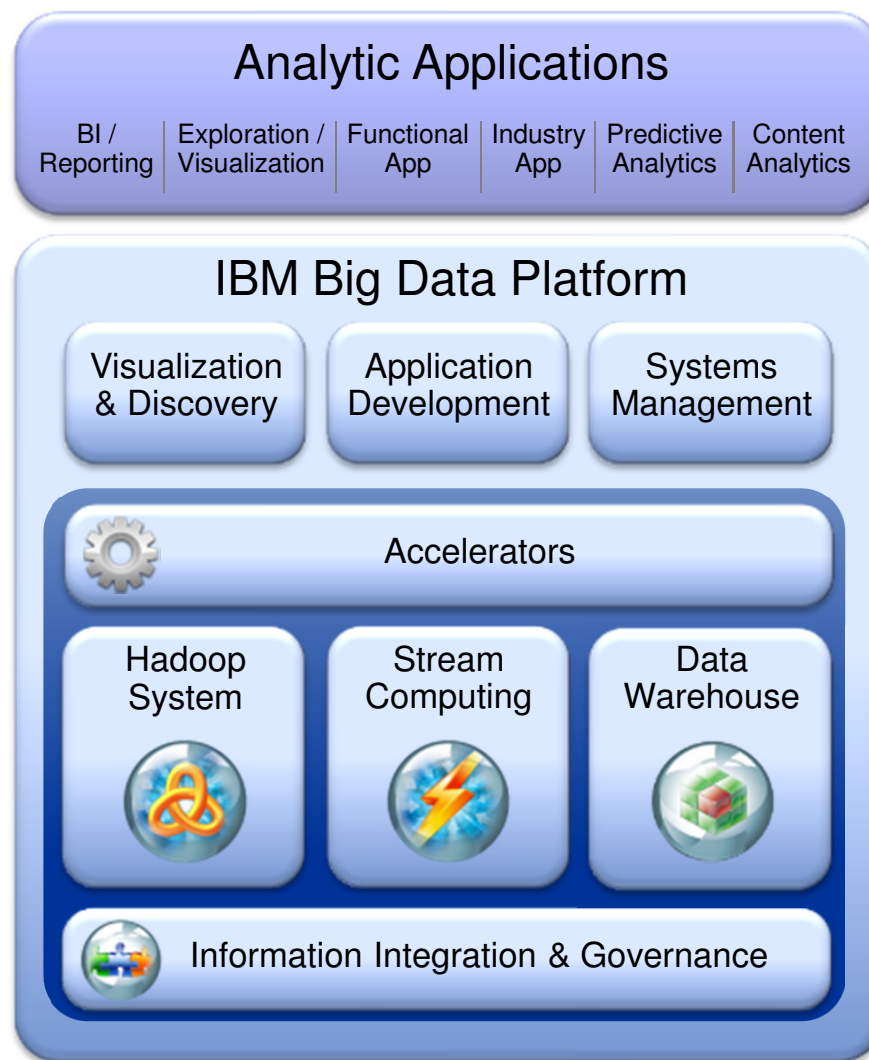
Exploring data – a sandbox for ad-hoc analytics



IBM Big Data Strategy: Move the Analytics Closer to the Data

New analytic applications drive the requirements for a big data platform

- Integrate and manage the full variety, velocity and volume of data
- Apply advanced analytics to information in its native form
- Visualize all available data for ad-hoc analysis
- Development environment for building new analytic applications
- Workload optimization and scheduling
- Security and Governance



InfoSphere BigInsights Brings Hadoop to the Enterprise

- **Manages a wide variety and huge volume of data**
- **Augments open source Hadoop with enterprise capabilities**
 - Performance Optimization
 - Development tooling
 - Enterprise integration
 - Analytic Accelerators
 - Application and industry accelerators
 - Visualization
 - Security
- **Provides Enterprise Grade Hadoop analytics**



IBM Significantly Enhances Hadoop



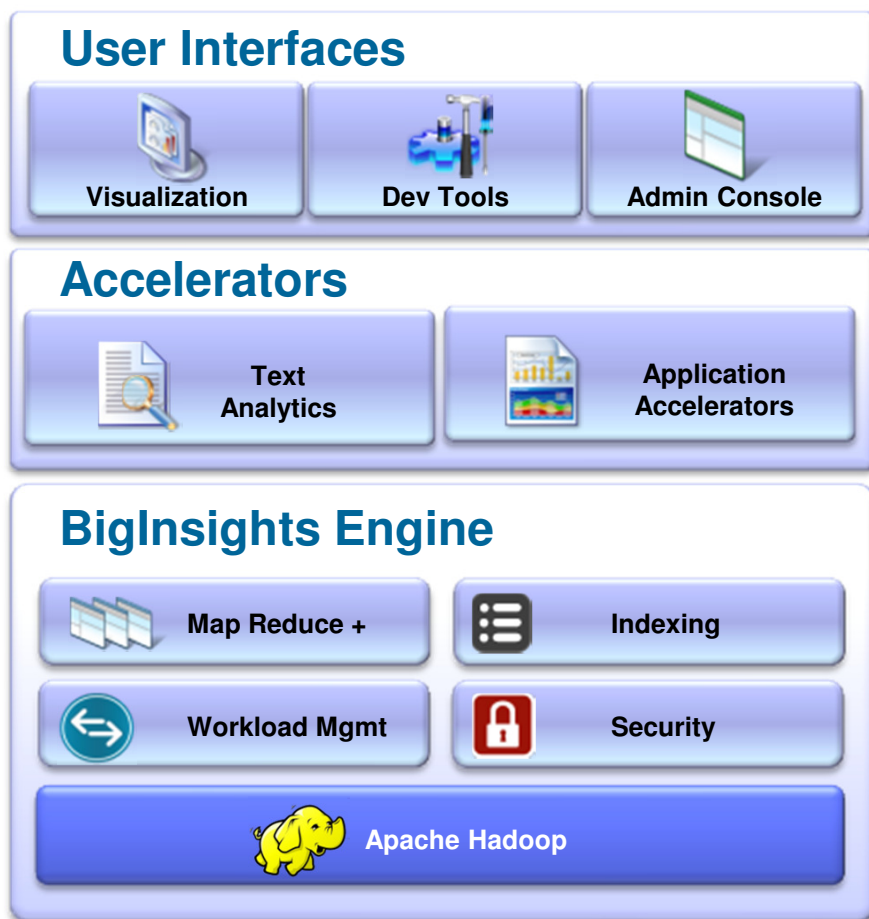
- **Scalable**
 - New nodes can be added on the fly.
- **Affordable**
 - Massively parallel computing on commodity servers
- **Flexible**
 - Hadoop is schema-less, and can absorb any type of data.
- **Fault Tolerant**
 - Through MapReduce software framework



IBM Innovation

- **Performance & reliability**
 - Adaptive MapReduce, Compression, Indexing, Flexible Scheduler
- **Analytic Accelerators**
- **Productivity Accelerators**
 - Web-based UIs
 - Tools to leverage existing skills
 - End-user visualization
- **Enterprise Integration**
 - To extend & enrich your information supply chain.

InfoSphere BigInsights – A Closer Look



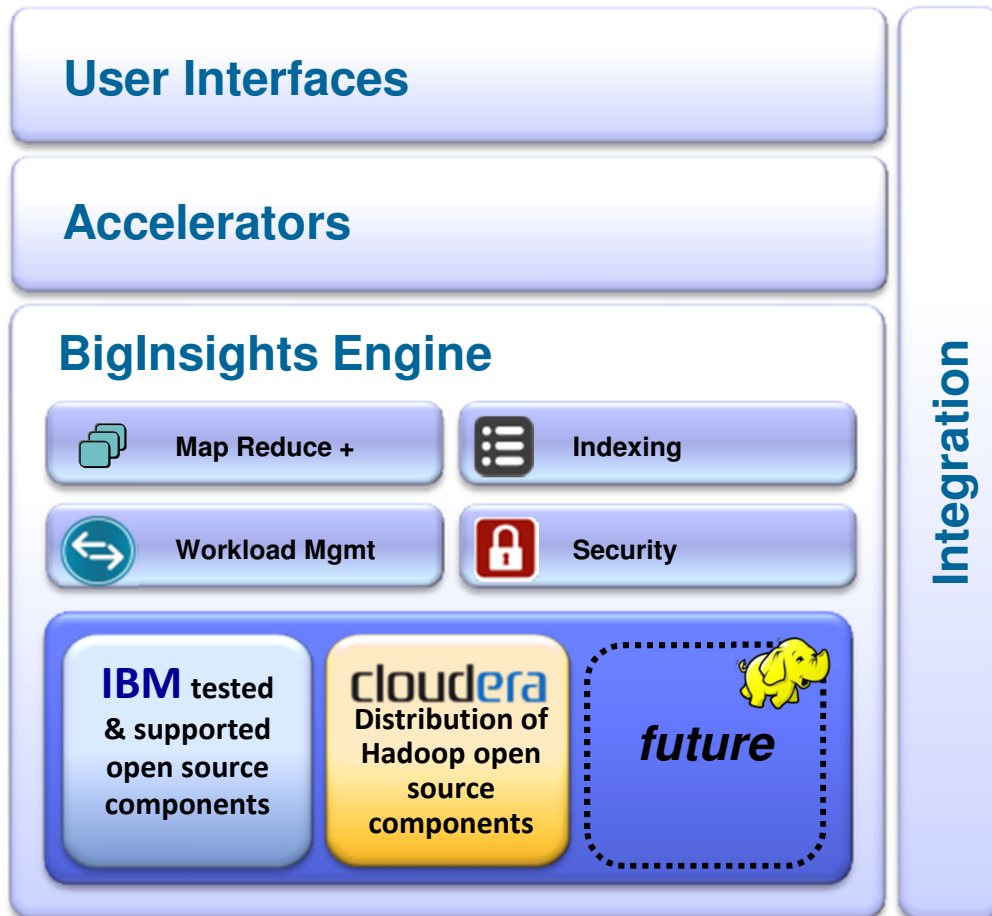
Integration



More Than Hadoop

- Performance & workload optimizations
- Unique text analytic engines
- Spreadsheet-style visualization for data discovery & exploration
- Built-in IDE & admin consoles
- Enterprise-class security
- High-speed connectors to integration with other systems
- Analytical accelerators

The Only Platform to Support Multiple Hadoop Distributions



- Provides a rich set of big data analytics and accelerators on top of open source
- Delivers a comprehensive big data platform on top of open source, that addresses all big data requirements.

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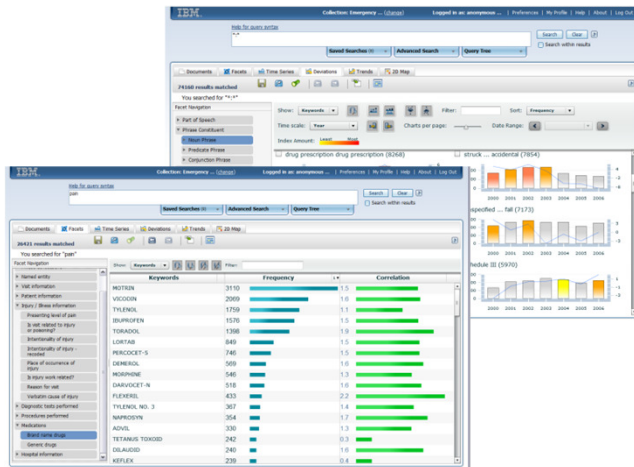
Text Analytics is the basis for Content Analytics

What is Text Analytics?

Text Analytics (NLP*) describes a set of linguistic, statistical, and machine learning techniques that allow text to be analyzed and key information extraction for business integration.

PC 143 (Hunter)
 15 June 2006 23:47
 Suspect identified himself as **John Setsuko**. Matched description given by night club doorman (IC1, Male, Ag 22-24 yrs, blue Everton shirt). Stopped whilst driving **White Ford Mondeo, W563 WDL**. Address given as **22 East Dene Ridge, Coppdock, Ipswich**. Searched at scene and found in possession of **1oz Cannabis Resin** and lockable pocket knife.

Arresting_Officer	PC 143
Arrest_Date_Time	15/06/2006 : 23:47
Suspect_Forename	John
Suspect_Surname	Setsuko
Suspect_VRN	W563WDL
Suspect_Vehicle_Color	White
Suspect_Vehicle_Make	Ford Mondeo
Suspect_Addr_Street	22 East Dene Ridge
Suspect_Addr_Town	Ipswich
Evidence_1_Description	1 oz Cannabis Resin
Classification	Drug possession



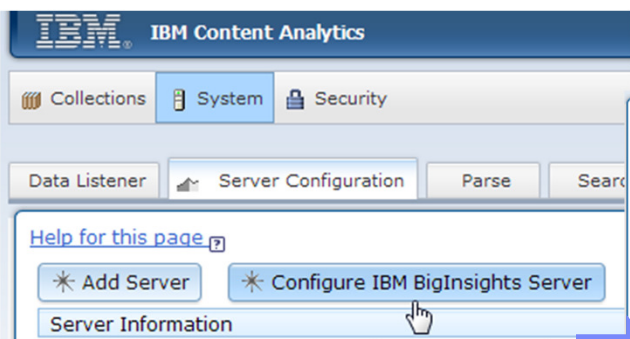
What is Content Analytics?

Content Analytics (Text Analytics + Mining) refers to the text analytics process plus the ability to visually identify and explore trends, patterns, and statistically relevant facts found in various types of content spread across internal and external content sources.

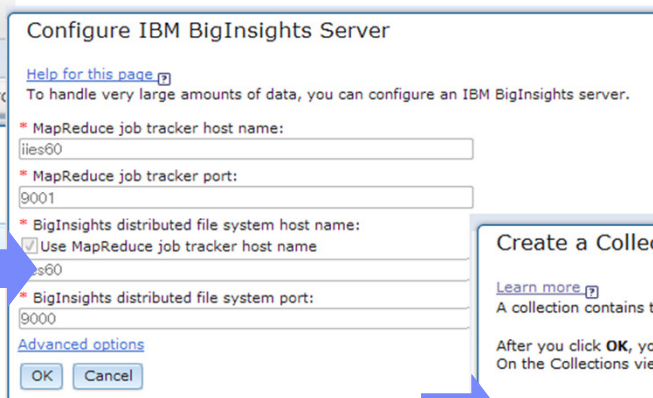
* Natural Language Processing

New in ICAwES v3.0 - Seamless Scale-out with BigInsights / Hadoop

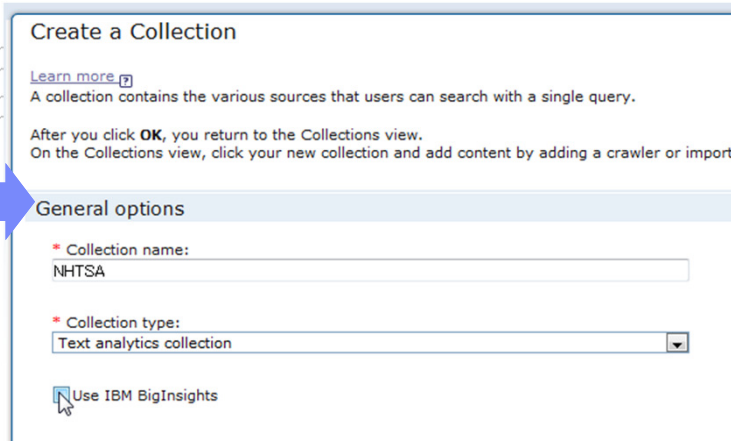
- Select "Configure BigInsights Server"



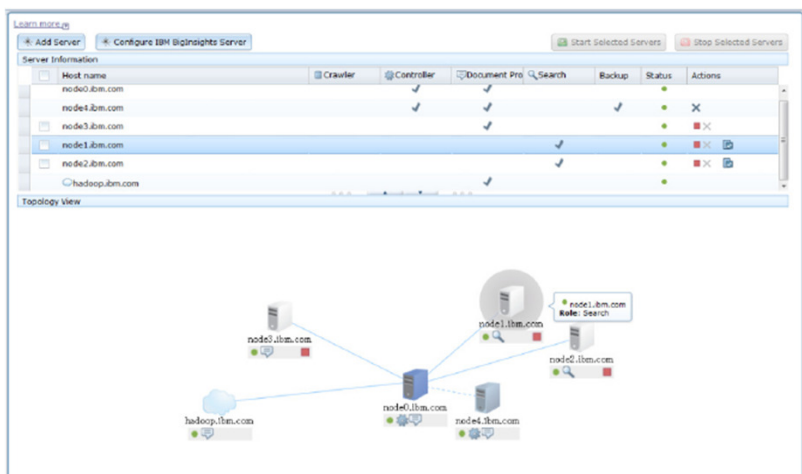
- Enter the BigInsights Server Information



- Specify "Use IBM BigInsights" as a Collection setting

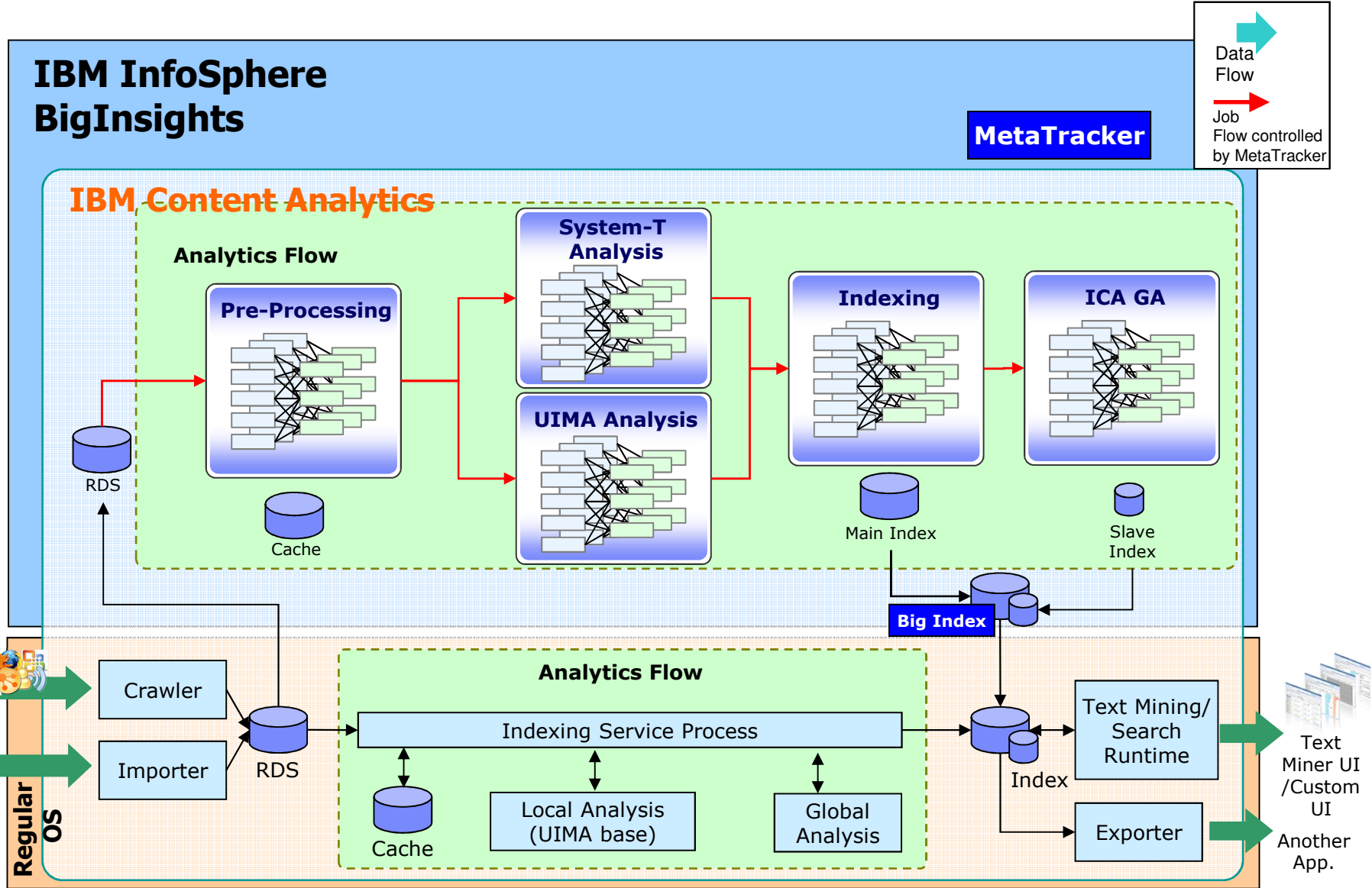


Admin user can confirm the setting on Topology View



...configuration files and ICA libraries, UIMA PEARs (including custom PEAR) and other required modules will be distributed to BigInsights servers automatically for that collection

New in ICAwES v3.0 - Seamless Scale-out with BigInsights / Hadoop



Documents View

The screenshot shows the IBM Content Analytics interface. The main window displays search results for 'BRAKE FAILURE' with columns for Flags, Source, Date, and Title. The results list several documents with their respective file paths and titles. A left-hand sidebar shows a Facet Navigation panel with various filters such as 'Part of Speech', 'Phrase Constituent', 'Named entity', 'Troubleshooting', 'Category for Auto', 'State', 'City', 'Vehicle/Equipment Corp', 'Vehicle/Equipment Make', 'Model', 'Model Year', 'Component Description', 'Date of Manufacture', 'Date of Purchase', 'Date of Fail', 'Anti-Lock Brakes?', 'Crash', 'Dealer's Name', 'Dealer's City', 'Dealer's State Code', and 'Dealer's Phone Number'. The main content area shows a list of documents with their titles and snippets of text, where 'BRAKE FAILURE' is highlighted in green.

- **Document Analysis Dialog**
 - Preview the whole document
 - Display document metadata and annotated facets with highlighted texts

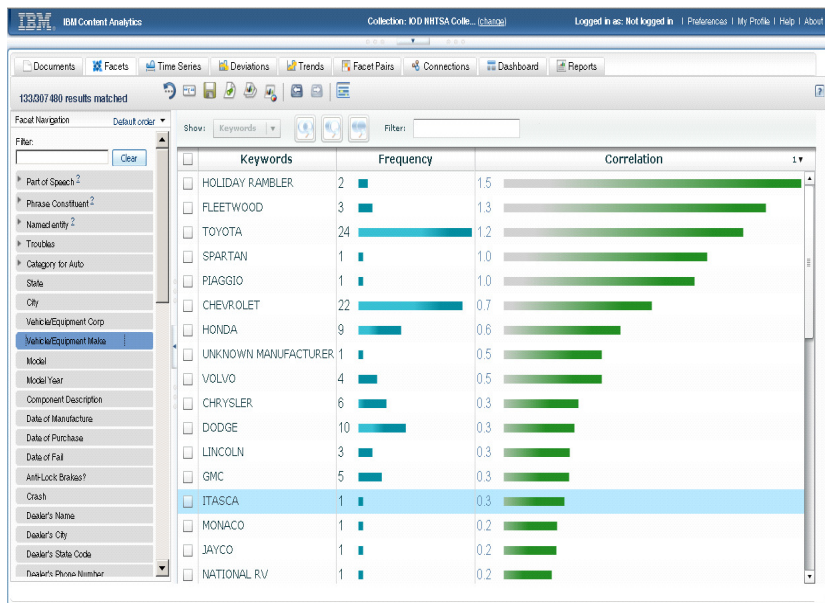
View the Documents based on current analysis criteria

- Flagging of documents
- Auto highlighting of key entities based on current query
- Export documents to be used in another system.

The screenshot shows the Document Analysis Dialog window. It displays a list of facets and their values. The facets are categorized into 'Analytics Facet' and 'Metadata Facet'. The 'Analytics Facet' list includes items like 'Noun - vehicle ... have', 'General Noun vehicle', 'Verb have', 'Verb be', 'Verb service', 'Preposition for ... parking', 'with Noun brake failure', 'Noun parking brake failure', 'Sequence failure', 'General Noun parking', 'General Noun brake failure', 'Modified Noun four ... time', 'adj four', 'Adjective four', 'General Noun time', 'Noun - vehicle ... have', 'Predicate vehicle ... have', 'General Noun vehicle', 'Verb have', 'Verb - Noun roll ... back', 'Verb roll', 'General Noun back', 'Preposition with Noun in ... park position', 'Noun park position', 'Sequence', 'General Noun park', 'General Noun position', and 'Noun - consumer ...'. The 'Metadata Facet' list includes items like 'Anti-Lock Brakes?', 'City', 'Component Description', 'Crash', 'Drive Train Type', 'Fire', 'Fuel System Code', 'Fuel Type Code', 'Vehicle/Equipment Make', 'Vehicle/Equipment Corp', 'Miles', 'Model', 'Number of Cylinders', 'Occurrences', 'Campaign Number', 'Original Owner?', and 'Reported to Police?'. The values for these facets are listed in a table format. A preview of the document text is shown at the bottom of the dialog, with 'BRAKE FAILURE' highlighted in green.

Metadata Facet	Name	Value
Anti-Lock Brakes?		Y
City		EAST PALATKA
Component Description		PARKING BRAKE
Crash		N
Drive Train Type		RWD
Fire		N
Fuel System Code		FUEL INJECTION
Fuel Type Code		GAS
Vehicle/Equipment Make		FLEETWOOD
Vehicle/Equipment Corp		FLEETWOOD ENTERPRISES, INC.
Miles		53000
Model		BOUNDER
Number of Cylinders		
Occurrences		1
Campaign Number		10095233
Original Owner?		N
Reported to Police?		N

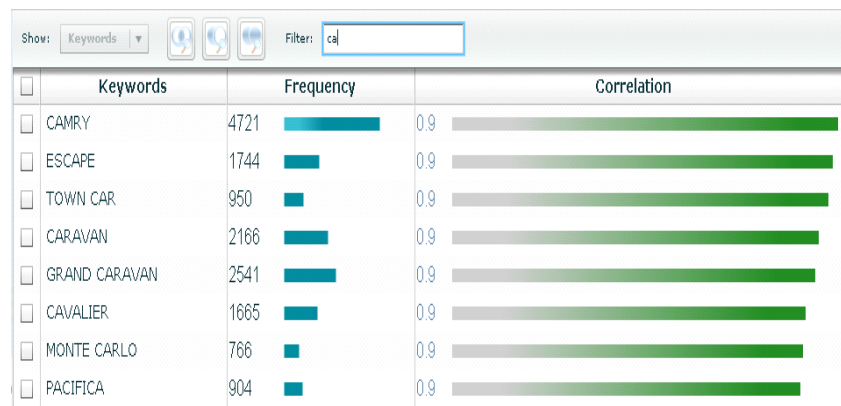
Facets View



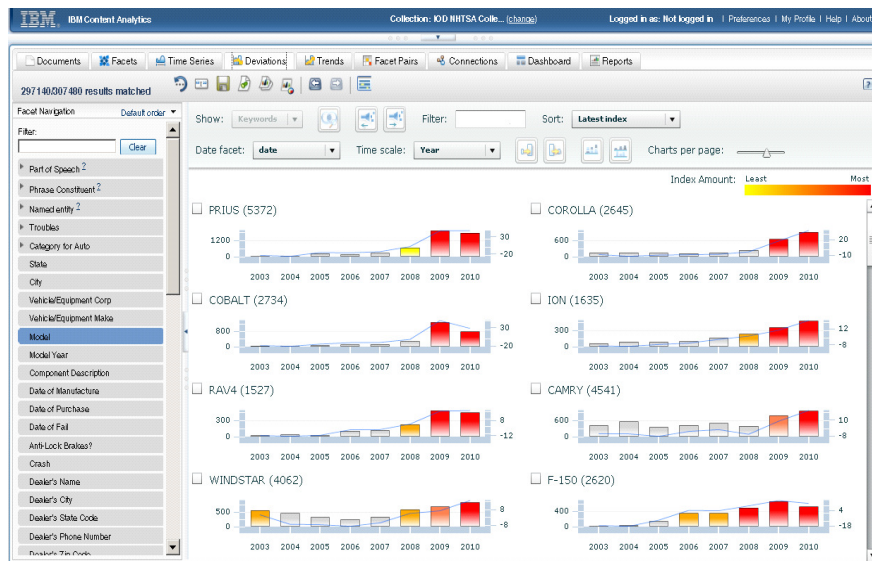
- Quick filter
- Multi-column sort

Show the frequency and correlation indices of keywords belong to the selected facet

- Add selected keyword to current search condition



Deviations View and Trends View

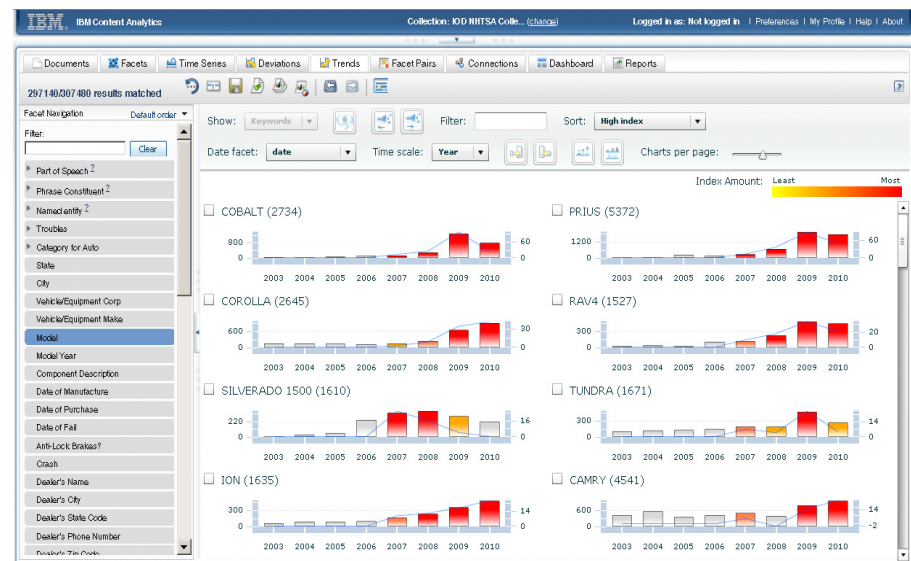


■ Trends View

- show how keyword in each time period deviates from average occurrence (against the whole current matched documents)..
- Predict pattern behaviors in future.

■ Deviations View

- show how keyword deviates from average occurrence between other keywords.
- Identify patterns that are cyclic.



Facet Pairs View

- Show the correlation between keywords belong to two different facets
- 3 view modes



Look at whole area

Table view

576752 results matched

Rows: Part of Speech	Columns: COMPONENT DESCRIPTION	Frequency	Correlation
transmission	POWER TRAIN:AUTOMATIC TRANS	28523	10.7
brake	SERVICE BRAKES, HYDRAULIC:ANT	27847	5.1
be	POWER TRAIN:AUTOMATIC TRANS	24898	1.1
and	POWER TRAIN:AUTOMATIC TRANS	23393	1.1
AK	SERVICE BRAKES, HYDRAULIC:ANT	19943	1.5
have	POWER TRAIN:AUTOMATIC TRANS	18711	1.2
be	ENGINE AND ENGINE COOLING:EN	18572	1.0
be	SERVICE BRAKES, HYDRAULIC:ANT	18230	0.9
and	SERVICE BRAKES, HYDRAULIC:ANT		
and	ENGINE AND ENGINE COOLING:EN		
tire	TIRES		
not	POWER TRAIN:AUTOMATIC TRANS	16362	1.1
vehicle	POWER TRAIN:AUTOMATIC TRANS	16245	1.1
engine	ENGINE AND ENGINE COOLING:EN	16044	4.7
vehicle	SERVICE BRAKES, HYDRAULIC:ANT	15273	1.2
AK	POWER TRAIN:AUTOMATIC TRANS	14806	1.0

Quick filter and sort

Grid view

VEHICLE SPE...	EQUIPMENT 4091	ELECTRICAL S...	SUSPENSION...	STEERING:W...	SUSPENSION...	SERVICE BRA...	ENGINE AND
54	737	379	334	208	273	215	314
5	2	1	0.9	0.6	0.7	0.6	0.9
25	702	326	467	218	188	222	262
9	2	0.9	1.3	0.6	0.5	0.6	0.7
39	688	306	384	201	293	180	297
8	2	0.9	1.1	0.6	0.8	0.5	0.9
38	515	344	348	251	454	291	349
1	1.5	1	1	0.7	1.3	0.8	1
18	339	339	231	3601	361	150	81
0.3	0.9	1	0.6	11.3	1.1	0.4	0.2
226	410	500	397	261	414	409	439
0.6	1.2	1.5	1.2	0.7	1.2	1.2	1.3
291	576	235	519				
0.8	1.7	0.7	1.6				
377	426	394	272	817	304	218	139
1.1	1.2	1.2	0.8	2.6	0.9	0.6	0.4
70	299	204	109	75	26	36	56
0.2	0.9	0.6	0.3	0.2	0.1	0.1	0.1

See in detail

New in ICAwES v3.0 – Sentiment Analytics

IBM Content Analytics with Enterprise Search
Collection: Sentiment Collection (change) | Logged in as: Not logged in | Preferences | My Profile | Help | About

637/637 results matched

Facet Navigation: Default order

Filter: [] Clear

Part of Speech²
Phrase Constituent²
Sentiment
My Keywords
product
prod_category
price

Show: Keywords | Filter: []

Sentiment	Values	Positive	Ambivalent	Negative
Admirable 6.5 cu. ft. Super Capacity Gas Dryer (34)		28	2	2
LuthorCorp 7.3 Cu. Ft. 7-Cycle Electric Dryer - White (18)		14	3	1
Extensive Enterprise 4.5 Cu. Ft. 14-Cycle Ultra Capacity High-Efficiency Washer - White (10)		6	2	2
Extensive Enterprise 4.5 Cu. Ft. 14-Cycle Ultra Capacity High-Efficiency Washer - Graphite Steel (10)		9	0	1
Extensive Enterprise SteamDryer 7.3 Cu. Ft. 14-Cycle Ultra Capacity Electric Dryer - Graphite Steel (10)		9	1	0
Extensive Enterprise 3.5 Cu. Ft. 7-Cycle High-Efficiency Washer - White (10)		8	1	0
EC 4.0 Cu. Ft. 26-Cycle King-Size Washer - Silver Metallic (10)		5	2	3
EC 4.0 Cu. Ft. 26-Cycle King-Size Washer - White (10)		0	4	6
EC 3.2 Cu. Ft. 9-Cycle Super Capacity Washer - White-on-White (10)		5	1	2

Document Preview | Trends

Analyze more in the Documents view

This product is good for the money. Works well and has a really large sized capacity.

There is nothing wrong with this dryer, as far as we can tell. It dries large loads in one go and has plenty of settings options.

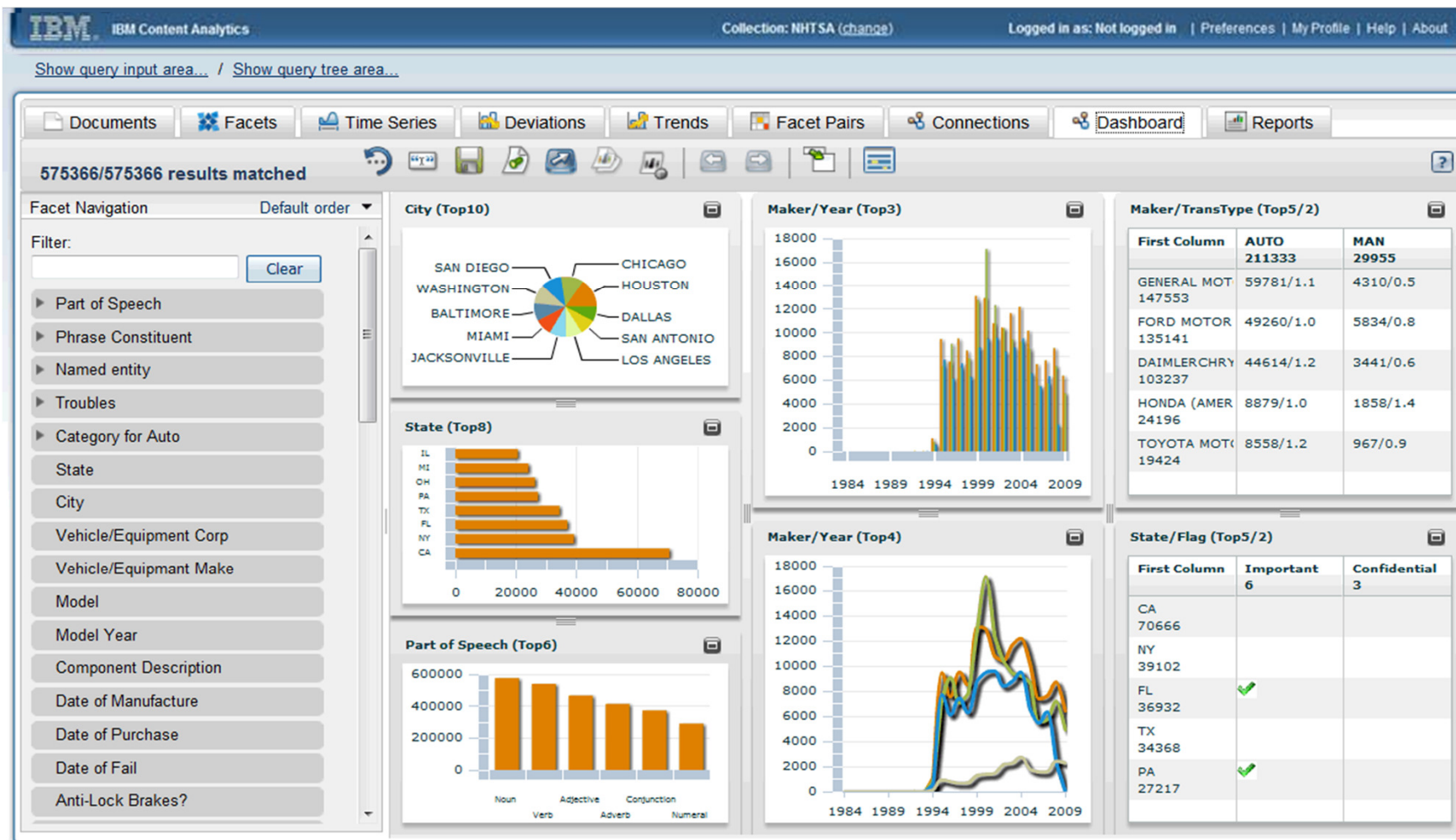
I was looking to buy similar Moreken products but couldn't justify the almost double cost for .5 cubic foot more. This dryer has been great. It fits plenty and dries clothes and bedding to completion without having to run a second cycle.

I was looking for n inexpensive dryer, and I needed to have it quickly. I never hear of Admirable

Sentiment Expressions

- Lists positive or negative expressions for selected facet value with colors
- Color represents the rank of correlation

Create Dashboard Views for Executive Summaries



[Click here](#)

IBM Content Analytics adds value to Big Data...



Healthcare Analytics

- **Analyzing:** E-Medical records, hospital reports
- **For:** Clinical analysis; treatment protocol optimization
- **Benefits:** Better management of chronic diseases; optimized drug formularies; improved patient outcomes



Customer Care

- **Analyzing:** Call center logs, emails, online media
- **For:** Buyer Behavior, Churn prediction
- **Benefits:** Improve Customer satisfaction and retention, marketing campaigns, find new revenue opportunities



Crime Analytics

- **Analyzing:** Case files, police records, 911 calls...
- **For:** Rapid crime solving & crime trend analysis
- **Benefits:** Safer communities & optimized force deployment



Insurance Fraud

- **Analyzing:** Insurance claims
- **For:** Detecting Fraudulent activity & patterns
- **Benefits:** Reduced losses, faster detection, more efficient claims processes



Automotive Quality Insight

- **Analyzing:** Tech notes, call logs, online media
- **For:** Warranty Analysis, Quality Assurance
- **Benefits:** Reduce warranty costs, improve customer satisfaction, marketing campaigns



Social Media for Marketing

- Analyzing:** Call center notes, SharePoint, multiple content repositories
- For:** churn prediction, product/brand quality
- Benefits:** Improve consumer satisfaction, marketing campaigns, find new revenue opportunities or product/brand quality issues



Interaction between Big Data and Content Analytics

- **IBM Content Analytics running on top of BigInsights for full analytics and rapid insights**
- **BigSheets and IBM Content Analytics – Filter with BigSheets then analyze in ICA for graphical view of analytic results**
- **IBM Content Analytics as a NLP engine to add define concepts and add depth to Cognos Consumer Insights**

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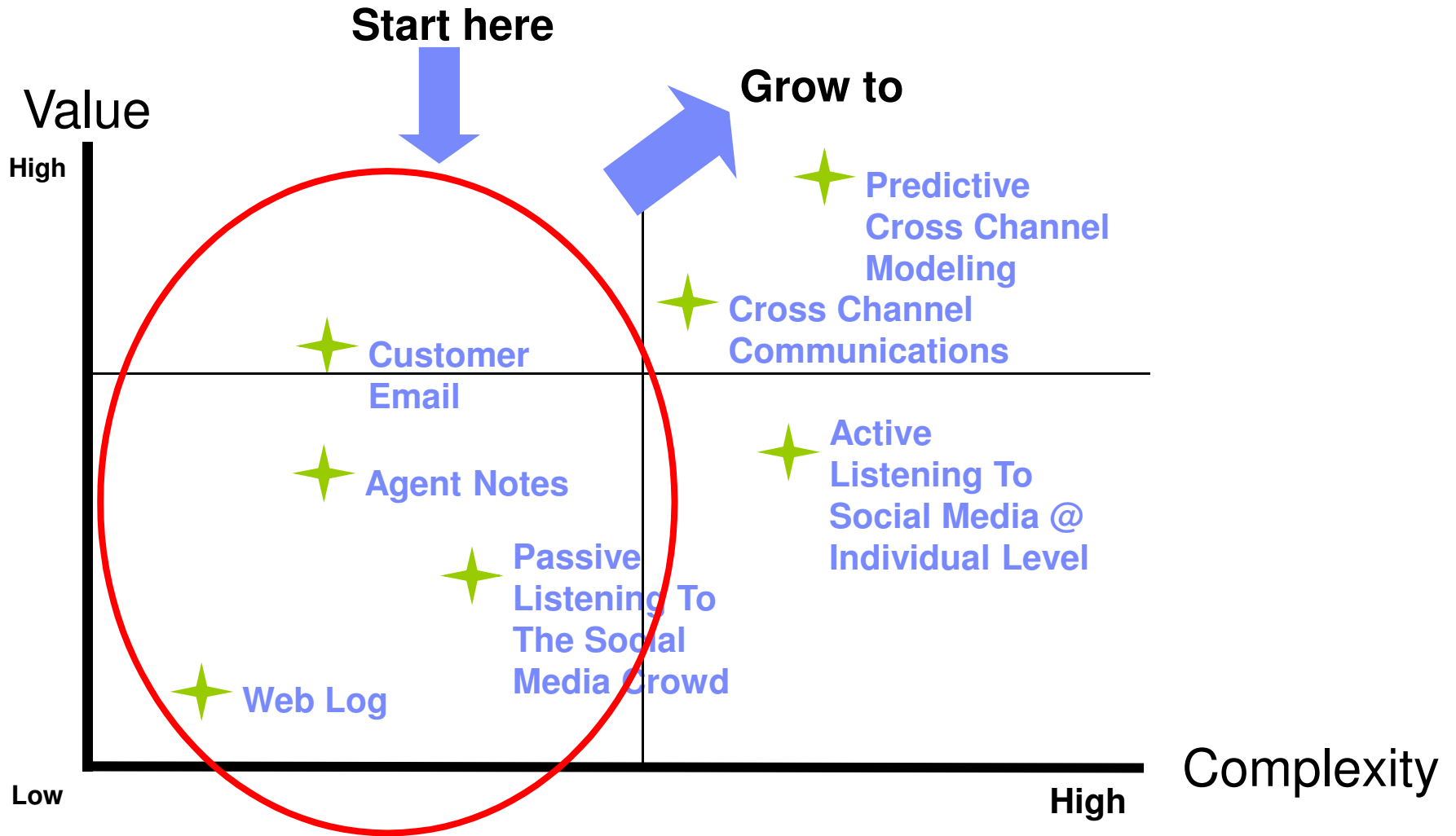
Thoughts On Getting Started



Use Case Selection Criteria

- Well proven path
- Can be done off-line/non-disruptive to existing systems
- Intuitive that there is low hanging fruit for additional insights
- Data set is already stored, but under-instrumented or overly summarized
- Initial findings can be arrived at 3 weeks or less
- Initial use cases have to be accretive to next set
- Initial use cases have to leverage common technology for next set of use cases

One Example Of How To Think About It



Use Case Starting Points

- **Mobile/Web Log**
 - Why: Reduce latency to under a day instead of weeks from 3rd parties
 - Why: Have the full data set for cross-correlation for other use cases
- **Agent Notes**
 - Why: Has proven to be an importance source of wisdom from the branches and Contact Centres
 - Why: An element of crowd sourcing
 - Why: Pulse of our internal teams whom we need to market to
- **Customer Email**
 - Why: They are often your best direct source of “pulse” of the business
 - Why: Lynchpin of crowd sourcing
 - Why: Six Sigma feedback component to cost take out
- **Passive Listening To The Social Media Crowd**
 - Why: Important early warning source of information
 - Why: You need the data – not summarization – for more advanced analytics
- **Cross Channel Communications**
- **Active Listening To Social Media @ Individual Level**
- **Predictive Cross Channel Modeling**

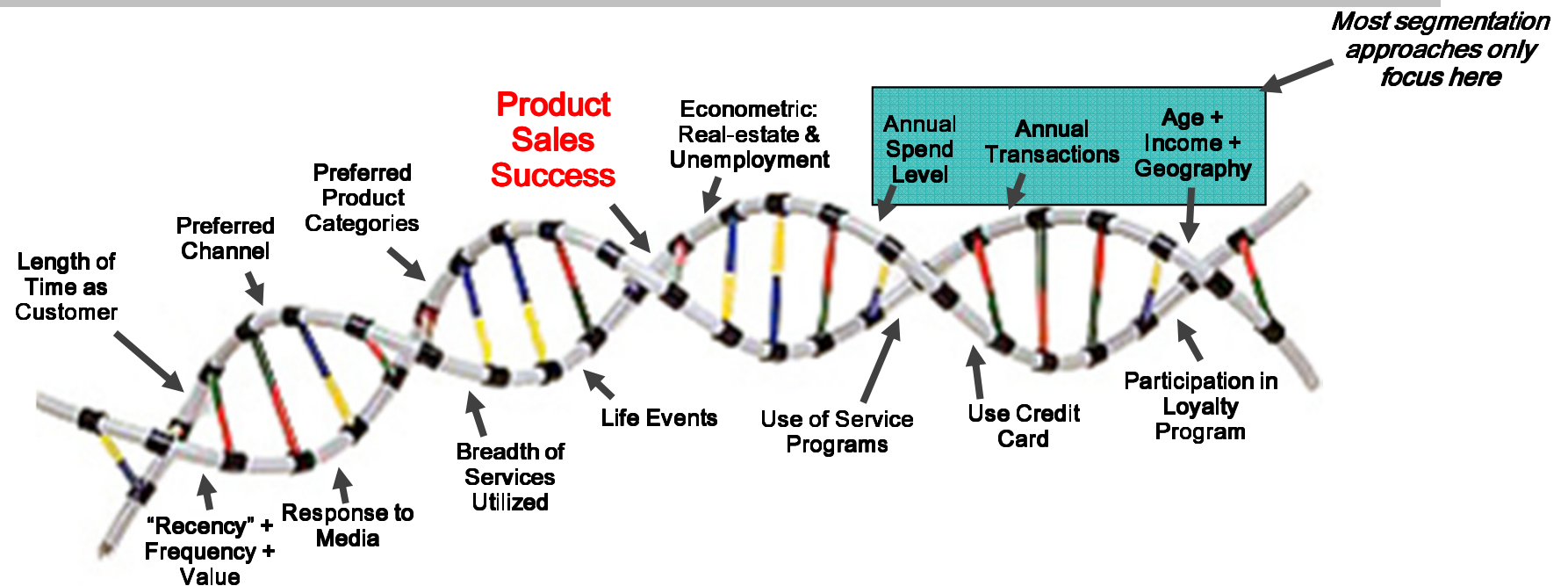
Examples Of Insight / Value Delivered

- **Web Log ***
 - Same day feedback on A|B testing new design (if sample is big enough)
 - Same day feedback on change in font/images/icons wording on behavior (if sample is big enough)
 - Input source into cross-channel behavior modeling
- **Agent Notes ***
 - Additional insight into customer “state” based on call handling
 - Feedback on features / policies not captured in formal system – nuance of customer opinion
 - Detecting of otherwise un-surfaced issues both at segment and individual customer level
 - Detection of staff dissatisfaction
 - Detection of otherwise hidden life events
- **Customer Email ***
 - Direct feedback on policy/fees
 - Notice of customers planning to attrition
 - Detecting of otherwise un-surfaced issues both at segment and individual customer level
 - (in private / brokerage situations early warning of law suits / advisor attrition)
 - Detection of otherwise hidden life events
 - Basis for cross-interaction behavioral / acceptance modeling
- **Passive Listening To The Social Media Crowd ***
 - Hour-by-hour feedback on announcements, offerings, policy changes
 - Taking the base reporting the rest of the org expects off the table so more important things can be focused on
 - Crowd feedback on competitor offerings
 - Path to SMARC

* All are expanded “feature vectors”

Case Study: Improvement Of Action Cluster Segmentation Approach

- Use of under-utilized data to expand modeled variables – “Feature Vectors” – The customers response to the firm’s value proposition
- Each feature vector is like a gene strand, which describes a facet, or set of customer behavior traits



Wait – How Do You Know If This Is Working?

- **Quick – how many of you were part of an experiment this week?**
- **Back testing**
- **A|B real world testing**
- **“Test Fast, Fail Fast, Correct Fast”**
- **Concept of MVP (minimum viable product)**

THINK

BIG

BIG