

Big Data Strategies with IMS

#16103 Richard Tran IMS Development <u>richtran@us.ibm.com</u>



Insert Custom Session QR if Desired.



SHARE is an independent volunteer-run information technology association that provides education, professional networking and industry influence.

Copyright (c) 2014 by SHARE Inc. C (i) (S) (i) C (i) C



Agenda



- Big Data in an Information Driven economy
- Why start with System z
- IMS strategies for big data
- Summary / Call to action





On a Smarter Planet, Unprecedented Changes are Occurring



- Business models under constant pressure
- Customers are more demanding and connected
- Great relationships trump great products



And leaders are responding by...









Providing a Great Experience

Offering Value In Every Interaction Innovating Across the Ecosystem





But what is Big Data?

Google can give you nearly 2 Billion options
 Vendors have even more definitions

What is Big Data									
Web	Images	Maps	Shopping	News					

About 1,940,000,000 results (0.18 seconds)

Here is how Gartner defines Big Data

Big data is high-volume, high-velocity and high-variety information assets that demand cost-effective, innovative information processing for enhanced insight and decision making.

Gartner research note "Survey Analysis - Big Data Adoption in 2013 Shows Substance Behind the Hype" Sept 12 2013

Analyst(s): Lisa Kart, Nick Heudecker, Frank Buytendijk

5





We've moved into a new era of computing - V^4



6

Agenda



- Big Data in an Information Driven economy
- Why start with System z
- IMS strategies for big data
- Summary / Call to action





The Big Data starting point

Types of Data Analysed

Transactional sources are the dominant data types analyzed in big data initiatives





Complete your session evaluations online at www.SHARE.org/Pittsburgh-Eval

Analyst(s): Lisa Kart, Nick Heudecker, Frank Buytendijk



The Big Data starting point Types of Big Data Analysed by Industry Transactional sources are the dominant data types analyzed in big data initiatives

	Manufacturing and Natural Resources	Media/ Communications	Services	Government	Education	Retail	Banking	Insurance	Healthcare	Transportation	Utilities
Transactions	73%	62%	67%	67%	54%	93%	83%	81%	75%	79%	80%
Log data	44%	57%	58%	59%	54%	40%	66%	61%	33%	71%	60%
Machine or sensor data	53%	38%	35%	33%	31%	27%	27%	48%	42%	50%	40%
Emails /documents	27%	43%	43%	41%	46%	27%	34%	39%	17%	29%	20%
Social media data	32%	52%	39%	26%	54%	73%	27%	13%	-	50%	-
Free-form text	17%	24%	28%	30%	31%	20%	34%	35%	67%	21%	40%
Geospatial data	27%	14%	19%	19%	38%	27%	27%	26%	8%	29%	40%
Images	19%	24%	17%	11%	38%	13%	5%	16%	25%	7%	•
Video	8%	29%	12%	7%	31%	13%	-	6%	8%	7%	•
Audio	10%	19%	8%	4%	8%	-	-	6%	-	-	-
Other	8%	14%	13%	15%	8%	7%	10%	16%	42%	14%	-
<i>n</i> =	59	21*	127	27*	13*	15*	41	31	12*	14*	5*

Note: Highlighted cells indicate the top three data types by industry. Multiple responses allowed

Source: Gartner (September 2013)

Gartner research note "Survey Analysis - Big Data Adoption in 2013 Shows Substance Behind the Hype" Sept 12 2013

Analyst(s): Lisa Kart, Nick Heudecker, Frank Buytendijk



The role of zEnterprise in Big Data analytics



10



- A large percent of the data that is accessed for analytics originates/resides on IBM zEnterprise
 - 2/3 of business transactions for U.S. retail banks
 - 80% of world's corporate data
- Businesses that run on zEnterprise
 - 66 of the top 66 worldwide banks
 - 24 of the top 25 U.S. retailers
 - 10 of the top 10 global life/health insurance providers
- 1,300+ ISVs run zEnterprise today, more than 275 of these selling over 800 applications on Linux
- The downtime of an application running on System z equates to approximately 5 minutes per year
- The System z mainframe can run over a thousand virtual Linux images on a single frame the size of a

omplete your session evaluations online at www.SHARE.F@frigeratori



Majority of today's analytics based on relational / "Structured" Data

- Analytics and decision engines reside where the DWH / transaction data is
- "Noise" (veracity) surrounds the core business data
 - Social Media, emails, docs, telemetry, voice, video, content
- What data are you prepared to <u>TRUST?</u>
- Where do you put your trusted Data?

"Circle of trust"





Demand for <u>differently</u> structured data to be seamlessly integrated, to augment analytics / decisions

- Analytics and decision engines reside where the DWH / transaction data is
- "Noise" (veractity) surrounds the core business data
 - Social Media, emails, docs, telemetry, voice, video, content
- Expanding our insights getting closer to the "truth"
 - Lower risk and cost
 - Increased profitability "Circle of trust" widens





Forward Thinking Organizations are Creating Value From Big Data

The power of Data coming together...



...with the power of Technology...





- 1. Enrich your information base with Big Data Exploration
- (a)
- 2. Improve customer interaction with Enhanced 360° View of the Customer



3. Optimize operations with Operations Analysis



4. Gain IT efficiency and scale with Data Warehouse Augmentation



5. Prevent crime with Security and Intelligence Extension





in Pittsburgh

Fraud Detection – Claiming disability allowance.



Data from Social Media

sites analyzed with Text

analytics



"Dude – awesome vacation"

Facebook Post



Complete your session evaluations online at www.SHARE.org/Pittsburg

Hadoop

or

agency

Result Set for

further processing

Result set uploaded or directly imported into OLTAP DBMS

Enterprise Integration and Governance... the key to success of incorporating Big Data

that come out







Watson Explorer

IBM BIG DATA PLATFORM Logical platform with many physical deployment options

Find, navigate, visualize all data Accelerators **BIG DATA PLATFORM** Speed time to value with analytic and application Systems Application accelerators Discovery Development Management InfoSphere BigInsights **Accelerators** Bringing Hadoop to the enterprise **InfoSphere Streams** Hadoop Stream Data Analytics for data in-motion exploration Warehouse System Computing **Information Integration & Governance InfoSphere Data Warehouse** Delivers deep insight with advanced database analytics & operational analytics Information Integration and Governance Governs data quality and manages the information lifecycle

Core data management solutions for the 21st Century



IMS 13



DB2 11 for z/OS

Unmatched availability, reliability, and security for business critical information



Delivering the highest levels of performance, availability, security, and scalability in the industry

- Up to 40% CPU reductions and performance improvements for (OLTP), batch, and business analytics
- Improved data sharing performance and efficiency
- Integration with InfoSphere BigInsights[™] / Hadoop and noSQL support
- Improved utility performance and additional zIIP eligible workload



- Breaking through 100k TPS 800% greater than IMS 12
- CPU reductions up to 62% for Java Apps
- SQL access to IMS data from both .NET and COBOL applications
- Greater flexibility and faster deployment for new applications with database versioning
- Big Data Ready exploitation of Hadoop / Big Insights, MDA, Watson Explorer...

IBM PureData System for Hadoop



Accelerate Hadoop analytics with appliance simplicity

Accelerate Big Data projects with built-in expertise

- Explore new ways to use all data
- Unlock new insights from unstructured data
- Establish a cost efficient on-line data archive Simplify with integrated system management
 - InfoSphere BigInsights software
- Compute and Storage hardware Ensure production grade security and governance Easily integrate with other systems in the IBM big data platform

CIO:Insight Apr 29 2013...Issues surrounding how long it takes to get a Hadoop application into production coupled with a lack of real-time capabilities are proving to be important barriers to deployment. overall Hadoop environment remain relatively small.



Simplify

Big Data for the

enterprise

Approaches

Processing done outside z (Extract and move data)



Processing in Appliance (z remains the master)

2





\$\$\$\$

Additional infrastructure. Challenges with scale, governance, ingestion.

\$\$

Appliance approach with PureData System for Hadoop. High speed load. z is the control point.

\$\$\$

Provision new node quickly Near linear scale. High speed load. z is the control potentiate

Z is the control point.



Data Explorer : visualization & discovery across all your data screes : "Integration at the glass

Securely connect to and leverage data stored in DB2 for z/OS & IMS

http://w

InfoSphere Data Explorer

Providing unified, real-time access and fusion of big data unlocks greater insight and ROI Help prioritize your System z big data integration and analytics projects

risk & ensure data

compliance

Improve customer service & reduce call times



Create unified view of ALL information for real-time monitoring



Increase productivity & Analyze customer information leverage past work & data to unlock true increasing speed to market customer value ²⁰ Complete your session evaluations online at www.SHARE.org/Pittsburgh-Eval

Agenda



- Big Data in an Information Driven economy
- Why start with System z
- IMS strategies for big data
- Summary / Call to action







- An open source software framework that supports data-intensive distributed applications
 - High throughput, batch processing
 - runs on large clusters of commodity hardware
 - Yahoo runs a 4000 nodes Hadoop cluster in 2008
- Two main components
 - Hadoop distributed file system
 - self-healing, high-bandwidth clustered storage
 - MapReduce engine



BIG DATA is not just HADOOP

23





Enhancing IMS analytics on System z with Big Data



- Much of the world's operational data resides on z/OS
- Unstructured data sources are growing fast
- There is a need to <u>merge</u> this data with trusted OLTP data from System z data sources
- IMS provides the connectors and the DB capability to allow BigInsights v2.1.2.0 to easily and efficiently access the IMS data source
 - BigInsights v2.1.2.0 is available on 3/13/2014



Enhancing IMS analytics on System z with Big Data



- Observation points lead to new business opportunities
- Observation points gleaned from both archived data and live data
- Score business events, track claims evolution, and more
- Make the data available to people who can do something meaningful with it

25





High level overview







n Pittsburgh

Import options BigInsights Platform HDFS IMS JDBC **POJO Import** Sqoop Import leve highnights Quick Start Edition (as a Header 000070.EVA / DPULASKI D21.7831.2003-05-26.MANAGER -3594.F 1985-07-09 -2022459.00 -1687590.56 -2 ter (Disease / Line 2 000090 EILEEN . WHENDERSON E11,5498 1971-05-15 MANAGER .3594 F 1985-07-09 -2022459 00.-16875 -136-10-156 3 000110.VINCENZO _ GLUCCHESSI A00.3490.1988-05-16 SALEREP _-3591.M 1985-07-09-2022459.00-1687 4 000120.SEAN __OCONNELL_A00.2167.1993-12-05.CLERK .-3596.M.1985-07-08-2022459.00-1687590.56 -* Properties file 5 000140.HEATHER , ARCHOLLS :COL 1783.2006-12-15.ANALYST ,-3593.F 1985-07-09.-1459.00.1687594 IGLECT + GRAN PRODUCT 6 000170 MASATOSHI ... JYOSHMURA. D11 2890 1995-09-15 DESIGNER. 3594 M 1985-07-2022459.00,-16 · DOLL IN CASE AND ADDRESS 7 000220.JENNIFER ... KLUTZ D11.0672.1998-08-29.DESIGNER - 3592 F 1985-07-09 - 2022459.00 - 1687596.5 000270, MARIA , LPEREZ , D21, 9001, 2006-09-30, CLERK , 3595 F 1985-07-08 - 2022 459 00 - 1687580 56 - 238 + things I because · Despei des + Little random day * Inclusion in such Application Highly

Sqoop Import



- Command line interface application for transferring data between RDBMS and HDFS.
- Import into Hive and Hbase
- Export from HDFS back into RDBMS
- Import:
 - Divides table into ranges using primary key max/min (can use split-by param)
 - Creates mappers for each range
 - Mappers write to multiple HDFS nodes
 - Creates text or sequence files
- Export:
 - Reads files in HDFS directory via MapReduce
 - Bulk parallel insert into database table.





Sqoop Import

- Import into HDFS using the below command:
- ./sqoop import --connect jdbc:ims://ecwas09.vmec.svl.ibm.com:5555/BIGDATP1 --driver com.ibm.ims.jdbc.IMSDriver --table EMPLOYEE -m 3 --split-by EDLEVEL --username 'OMVSADM' -P
- 13/06/19 17:50:27 INFO db.DataDrivenDBInputFormat: BoundingValsQuery: SELECT MIN(EDLEVEL), MAX(EDLEVEL) FROM EMPLOYEE
- 13/06/19 17:50:46 INFO mapreduce.ImportJobBase: Transferred 5.123 KB in 20.3762 seconds (257.4572 bytes/sec)
- 13/06/19 17:50:46 INFO mapreduce.ImportJobBase: Retrieved 43 records.
- HDFS Output (below)

Header

000070,EVA, , DPULASKI, D21,7831,2003-05-26, MANAGER, -3594, F, 1985-07-09, -2022459.00, -1687590.56, -23 1 000090, EILEEN , WHENDERSON , E11, 5498, 1971-05-15, MANAGER , -3594, F, 1985-07-09, -2022459, 00, -168759 2 000110,VINCENZO, GLUCCHESSI A00,3490,1988-05-16,SALEREP .-3591,M,1985-07-09,-2022459.00,-16875 3 000120.SEAN ... O'CONNELL .A00.2167.1993-12-05.CLERK .-3596.M.1985-07-09.-2022459.00.-1687590.56.-2: 4 5 000140.HEATHER __ANICHOLLS .C01.1793.2006-12-15.ANALYST .-3593.F.1985-07-09.-2022459.00.-1687590. 000170, MASATOSHI, JYOSHIMURA, D11, 2890, 1999-09-15, DESIGNER, -3594, M, 1985-07-09, -2022459.00, -168 6 000220, JENNIFER , , KLUTZ , D11,0672,1998-08-29, DESIGNER, -3592, F, 1985-07-09, -2022459.00, -1687590.56 7 000270, MARIA . . LPEREZ .D21, 9001, 2006-09-30, CLERK .-3595, F, 1985-07-09, -2022459.00, -1687590.56, -2352 8



SHARE,

BigInsights Database Import Application

 Utilize the built in Database Import Application by providing the database connection parameters;

eicome Dashboard Cluster Si	tatus Files Applications App	lication Status BigSheets
n Manage Link		
pplications	+ Execution	
earch	Execution Name: CognosDB	► Run
	✓ Parameters	
R	* Properties file:	/user/biadmin/credstore/imsdb.properties
		SELECT * FROM PRODUCT
	* SQL statement:	
Database Import	4	
	*Output format:	csv
	Output directory:	/user/biadmin
	CSV delimiter:	comma
	* Include Column Headers:	

30



BigInsights Database Import Application

• Once the run is completed, view the data in HDFS:

lication History										E
Status	Execution Name	e	Pro	ogress		Start Time		apsed Time (sec)	Output	De
No filter applied										
		1	100%	2	2013 10 3 20:53:20		14			
	CognosDB			100%	2	2013 10 3 20:41:22		18	B	
Welcome Dashboan	d ClusterStatus Fil	es Applicatio	ons Applicati	on Status Big	Sheets					
		Path: /user/bia	idmin/output.txt					T Go		
HDFS		-	Name	Size	Block Size	1	Time	Pe	mission	Own
hdts://biven:9000/		Output to 7.8 KB 1.28 0 MB Old 3, 2013 853 31 PM IW + - + - biadmin Edit Viewing Size: 10KB - Text @ Sheet -								
Agringits Adoop Adoop		hdfs://bivm.9 Comma Separ	000/user/bindmin/ ated Value (CSV) D	butput.txt Data 🤌 💽	Manter Workbook					
		Ready							2 Re	streish
			GOSALES_RO	OTKEY	COUNTRY_CO	ODE C	OUNTRY_EN		FLAG_IMAGE	
		1 1			1.003	United States		F03 jpg		
applications		2 1			1004	Canada		F04 jpg		
🖃 😂 bladmin		3 1			1020	Mexico		F19.jpg		
💽 🗀 .staging		4 1			1021	Brazil		F20.jpg		
Concedistore	6	5 1			4011	Japan		₽11jpg		
(I) Con Secolar bits		6 1			4012	Singapore		F12 jpg		
a corrie-bid	82	7 1			4013	Korea		F13.jpg		
e e obre bie		Transa and the second sec								
output tet		8 1			4014	China		F14.jpg		



Complete your session evaluations online at www.SHARE.org/Pittsburgh-Eval

31



BigInsights BigSheets

This data can be saved as BigSheets workbook for further analytics

1	BM InfoSp	here BigInsight	ts		Welcome biadm	iin L	.og out A	bout Information (Center	IBM.
	Welcome	Dashboard (Cluster Status	Files App	lications Ap	plicat	ion Status	BigSheets		
Wo	rkbooks > Vie	w Results								23
-	nlavan A									
Em	ployee Ø		1	2 ¹⁰ -1						
×	Delete	Add chart - II Chi	Build new	workbook						
💟 F	Ready			2 Refresh	🖽 Fit column(s)	E E	Export as 💌	🕞 Run 🔳 Stop	100%	% 👻
		EMPNO	+	FIRS	TNME	-		MIDINIT	*	
1	10		CHRIS	TINE		1				HAAS
2	20		MICHA	AEL		1				THON
3	30		SALLY	1		1	4			KWAN
4	50		JOHN			1	В			GEYE
5	60		IRVING	3		1	ŧ.			STER
6	70		EVA			1	D			PULA
7	90		EILEEI	N		1	N			HEND
8	100		THEO	DORE		(2			SPEN
9	110		VINCE	NZO		1	G			LUCC
0	120		SEAN							0'00
11	130		DELO	RES		1	M			QUIN
2	140		HEATH	IER			Δ.			NICH
3	150		BRUC	E						ADAM
14	160		ELIZAE	BETH		1	R			PIANE
15	170		MASA	TOSHI			100			YOSH
16	180		MARIL	YN			5			scol
17	190		JAMES	õ		4	H			WALK
18	200		DAVID							BROV
19	210		WILLIA	M		1	T			JONE
20	220		JENNI	FER		1	ĸ			LUTZ
21	230		JAMES	5			rg			JEFFE
22	240		SALVA	TORE		1	M			MARI
23	250		DANIE	1			5			SMIT



Elevated demand for business analytics drives new requirements and focus



Enterprise-level scale & performance Mission critical availability Faster access to operational data Rapid, cost effective deployment & expansion More integrated view of data across the environment

Modernization

- Standardization & Consolidation
- Operational BI
- Data Governance
- Cloud Computing





Watson Explorer

 Watson Explorer is the visualization & discovery capability for IBM's comprehensive big data platform

 Watson Explorer is a key component of all the big data use cases with greatest impact in Big Data Exploration & Enhanced 360 View of the Customer





Watson Explorer : visualization & discovery across all your data sources : "Integration

Securely connect to and leverage data stored in DB2 for 7/0S & IMS

http://w

Watson Explorer

Providing unified, real-time access and fusion of big data unlocks greater insight and ROI

Help prioritize your System z big data integration and analytics projects

risk & ensure data

compliance

Improve customer service & reduce call times

at the glass

Create unified view of ALL information for real-time monitoring



Analyze customer information Identify areas of information & data to unlock true

RISK

leverage past work 35 Complete your sessing speed to market ARE.org/Picustomer value

Increase productivity &



Watson Explorer product architecture and differentiators



Differentiators

Federated discovery and navigation

Scalable architecture

Accurate results

Secure connectivity

Powerful development tools

Unique application framework

Fast time to value





Seamless IMS integration





IMS + Watson Explorer -Configuring the IMS source

After deploying the IMS JDBC driver, create a new Database seed

Seed Component: Database (Custom SQL)	
Host	ec01255.vmec.svl.ibm.com
Port	5,555
Username	omvsadm
Password	****
Database system	IMS
Database name	BMP355
SQL Statement	SELECT * FROM PCB01.HOSPITAL, WARD, PATIENT
Key Column	PATNAME
Advanced Configuration (5)	
JDBC Connection String	jdbc:ims://ec01255.vmec.svl.ibm.com:5555/BMP355:dpsbOnCommit=true;
JDBC Class	com.ibm.ims.jdbc.IMSDriver



Complete your session evaluations online at www.SHARE.org/Pittsburgh-Eval

38



IMS + Watson Explorer -Setting up the data transformation

 After creating a new seed, a converter needs to be configured using standard XPATH

Converter Component: Database seeds support	
Тур	e-In application/vxml-db
Туре	Out application/vxml-db
Fallt	vack (unset)
Output for	king (unset)
Ν	ame





Original IMS hierarchy for hospital database

Hierarchy: HOSPITAL->WARD->PATNAME

Goal: Get a patient centric view





Why use Watson Explorer

- Previously to change the schema so that the PATIENT information is at the top, a logical database needs to be created
- This requires a DBA to be involved and a time window when the new database resources can be brought online
- Watson Explorer allows indexes to be created dynamically and for better searching that is not restricted to IMS Segment Search Arguments (SSAs)





Searching the IMS database with Watson Explorer

• Query: Who are the patients in the Alexandria hospital

IBM InfoSphere Data Explorer	Alexandria	Search
Topic Clusters	1. BOB DAVIS new wir	ndow preview
Not enough text to cluster	Hospital_hospcode:	R1210010000A, R1210010000A, R1210010000A
	Ward_wardno:	0004, 0004, 0004
	Hospil:	900
	Hospcode:	R1210010000A
	Hospname:	ALEXANDRIA
	Wardil:	900
	Wardno:	0004
	Wardname:	SURGICAL
	Patcount:	
	Patil:	900
	Patnum:	0001
	ims://ec01255.vmec.svl.it	om.com.5555/_ey-val=BOB DAVIS - 2K - cache - IMS_BMP355_EC01255
	2. KEVIN HITE new w	indow preview
	Hospital_hospcode:	R1210010000A, R1210010000A, R1210010000A
	Ward wardno:	0004, 0004, 0004
	HospII:	900
	Hospcode:	R1210010000A
	Hospname:	ALEXANDRIA
	Wardil:	900
	Wardno:	0004
	Wardname:	SURGICAL
	Patcount:	m
	Patil:	900
	Patnum:	0002
	ime //oc01255 ymoc cyl i/	om com 5555/ ov.val=KEVIN HITE - 2K - cacha - IMS_BMP355_EC01255





Searching the IMS database with Watson Explorer

• Query: Who are the patients currently in dermatology

IBM InfoSphere							
Data Explorer	Dermatology						
Topic Clusters	1. <u>VVILLIAIVI LI</u> new wir	ndow preview					
Not enough text to cluster	Hospital_hospcode:	R1210020000A, R1210020000A, R1210020000A					
5	Ward_wardno:	0002, 0002, 0002					
	HospII:	900					
	Hospcode:	R1210020000A					
	Hospname:	SANTA TERESA					
	Wardll:	900					
	Wardno:	0002					
	Wardname:	DERMATOLOGY					
	Patil:	900					
	Patnum:	0001					
	ims://ec01255.vmec.svl.ib	om.com:5555/_ev-val=WILLIAMTLL-2K-cache					





Machine Data Analytics Accelerator IT use cases:



IMS and IBM Accelerator for Machine Data Analytics



- Consume log data produced by Transaction Analysis
 Workbench
- Index and link transactions together across products (IMS, DB2, MQ, CICS, WebSphere)
- Make large amounts of IMS transactional log data available to the suite of BigInsights tools.







at all IMOCOVA

Transaction Analysis Workbench - Log conversion

na second in the second second second second in the second s

uaus	and 🔄 messagacometrici 🔄 wooacomitrinenticine	nnnencepionjava	Passworki E merva		unit Bunandaraa		
1	TIME;TranCode;Userid;RecToken;IMS	ID;InputQ;	Process;TotalTm;	CPUtime;FFGets;F	FUpdats;FPCal	ls;FPGets;FPUpdats;TP	ESAF;ESAFName
2	2013-08-11 22:01:38.244507;CCUTII	;DFSMTCN	T;C9D4E2C6404040	400 <mark>00100ba000</mark> 000	00;I <mark>MSF</mark> ;0	.000223;5.770438;5.77	066 <mark>1</mark> ;0;;;;;;;DB2A
3	2013-08-11 22:01:44.465656;CCUTII	;DFSMTCN	T;C9D4E2C6404040	40000 <mark>1</mark> 00BB000000	00; <mark>IMSF</mark> ;0	0.009176;0.041509;0.05	0685;0;;;;;;;DB2A
4	2013-08-11 22:04:59.984936;TSSIM	;	;C9D4E2C6404040	40000 <mark>100BC000000</mark>	00; <mark>IMS</mark> F ;8	0.716070;0.536743;9.25	2959;0 <mark>;;;;;;;</mark>
5	2013-08-11 22:05:09.223175;AUTOQR	Y ;	;C9D4E2C6404040	40000100BD000000	00;IMSF ;0).044868;0.638912;0.68	3318;0;;;;;;;DB2A
6	2013-08-11 22:06:01.494244;PDMSG2	; PDBMPLE	;C9D4E2C6404040	40000100c0000000	00;IMSF ;2	24.965797;0.467127;25.	432924;0;;;;;;;;
7	2013-08-11 22:06:26.796034;PS61	; PDBMPLE	;C9D4E2C6404040	40000100c1000000	00;IMSF ;0).130944;0.033681;0.03	5123;0;;;;;;;
8	2013-08-11 22:06:26.796171;PD32	; PDBMPLE	;C9D4E2C6404040	40000 <mark>100c2000000</mark>	00;IMSF ;0).142469;0.952377;0.96	7167;0;;;;;;;DB2A
9	2013-08-11 22:06:27.595731;EMAILE	MP; PDBMPLE	;C9D4E2C6404040	40000000000000000000000000000000000000	C4;IMSF ;0).302760;1.535334;1.54	5423;0;;;;;;; <mark>,</mark> QY2A
10	2013-08-11 22:06:23.509104;TSSIM	;	;C9D4E2C6404040	40000100BE000000	00;IMSF ;1	011752;0.269375;1.28	1127;0;;;;;;;

I management and the second second





Machine Data Analytics Accelerator - Watson Explorer Search

48

Home >		
Log Type	- Results for: - (2924 results, 0.067 seconds) 7 -	
+ Csv (2.924)	2013-08-11722:01:38-07:00 0.DB2A.IMSF.0.000223.5.770438.C9D4E2C6404040000100B400000000.2013-08-11.22:01:38.244507.5.770661.CCUTIL DFSMTCNT 284182159d4115f39d277118180e7765	
Batch ID + Imsmq (2,924)	Add Tag 2013-08-12102:08:18-07:00 0.IMSF_0.002689.0.094098.C9D4E2C64040400001029B00000000,2013-08-12 02:06:19.646477.0.097343,TSSIM, 437c10379fatc518dccbd0741020207a	
	Add Tag 2013-08-12T02:06:19-07:00 0.DB24.IMSF0.021285.0.263024.C9D4E2C5404040400001029C00000000.2013-08-12.02:06:19.722494.0.263668.AUTOORY	
	a31e9e117248507ea15de6520d21037b	
	Add Tag 2013-08-12102:06:27-07:00 0.IMSF .0.000758.0.948535.C9D4E2C5404040400001029D00000000,2013-08-12.02:06:27.794712.0.949293.PDMSG2 PDBMPLE d79ca026b062a58744b01t58d2a9cacc	
	Add Tag 2013-08-12T02:06:28-07:00 0,IMSF.0.171319.0.275632.C9D4E2C64040400001029E00000000,2013-08-12.02;06:28,561491,0.278735,PS61.PDBMPLE 5de1c08f0dad39453cbs2a54aff3d065	
	Add Tag 2013-08-12T02:06:37-07.00 0,IMSF.0.024757.0.342069.C9D4E2C64040400001029F00000000,2013-08-12 02:06.37.164637.0.366992.PCCBRSY2.X0OPSMVS 5c5858b0b0e039c90005c56d0f7ii92fb	
	Add Yag 2013-08-12T02:07:01-07:00 0,IMSF 0.461442.1.557907.C9D4E2C6404040000102A00000000,2013-08-12 02:07:01.871772.2.019529.PCCBRSY2.X00PSMVS 2ca0e9535f83699e6ecd28669b5be3a3	
	Add Tag 2013-08-12T02:07:21-07:00 0.IMSF 0.003015.0.003173.C9D4E2C840404040000102A100000000.2013-08-12.02.07:21.855832.0.006188,TSSON , *******	



Machine Data Analytics Accelerator – Data Analytics using BigSheets

IBM InfoSphere BigInsights Enterprise Edition									Welcome gluson Log	out About In	formation Center IE
Wekone	Dashboard	Cluster Status	Fles	Applications	Application Status	BigSheets					
lorkbooks >	View Results										
4ST250 0	,										
Delete	Add dont -	IMST250 : 001	dnew workbo	1							14
Ready										Run	5top 300%
L en a			-	-	<						
1.029			5		1						
1028 ·			/								
1.027 -			1								
026											
L024 -		1									
.023 -											
.022 -	1				1						
021		-									
1.02 -											
1.018											
.017											100
016											/
.005											
024											
013											
011	- 1										
.01											
009											
006											- 1
007											
005									~		
.004											
1.003					1						
0.002					1					-	
1001	2	2	2	2 2	2	2 5 4	2	2 2	a a	2 2	
50	8	13 B	3	24	5 3	8 8 9	33	22 03	\$ 8	50 B	8 B
											In Pätsburgh 2

Agenda



- Big Data in an Information Driven economy
- Why start with System z
- IMS strategies for big data
- Summary / Call to action



Conclusion / Call to action:

- For additional information including whitepapers and demos, please visit:
 - IBM big data for z web site
 - Smarter Computing
- Education

5

- Further developments:
 - Future webcast and announcements
 - World of DB2 for z/OS

Wanting to experiment on a big data integration project ? Partner with IBM Silicon Valley Laboratory. (richtran@us.ibm.com)

Develop your own big data strategy –Contact your local IBM sales representative to get started.







Thank You!

Your feedback is important to us

