

Collection and Interpretation of Big Data

YPSW AIHA Annual Meeting

Total Worker Exposure January 25th, 2019

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The business of sustainability

Data is the new asset



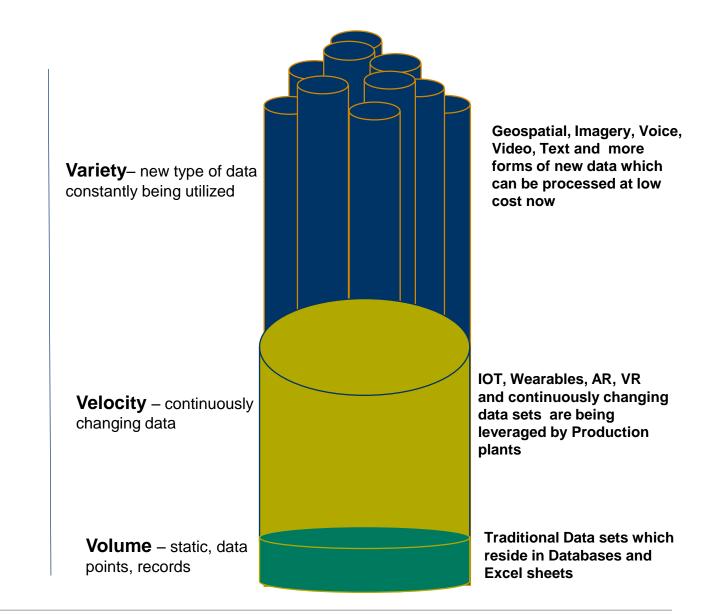
001478-3278 2017

Ten years on: banking after the crisis South Korea's unfinished revolution Biology, but without the cells

The world's most valuable resource

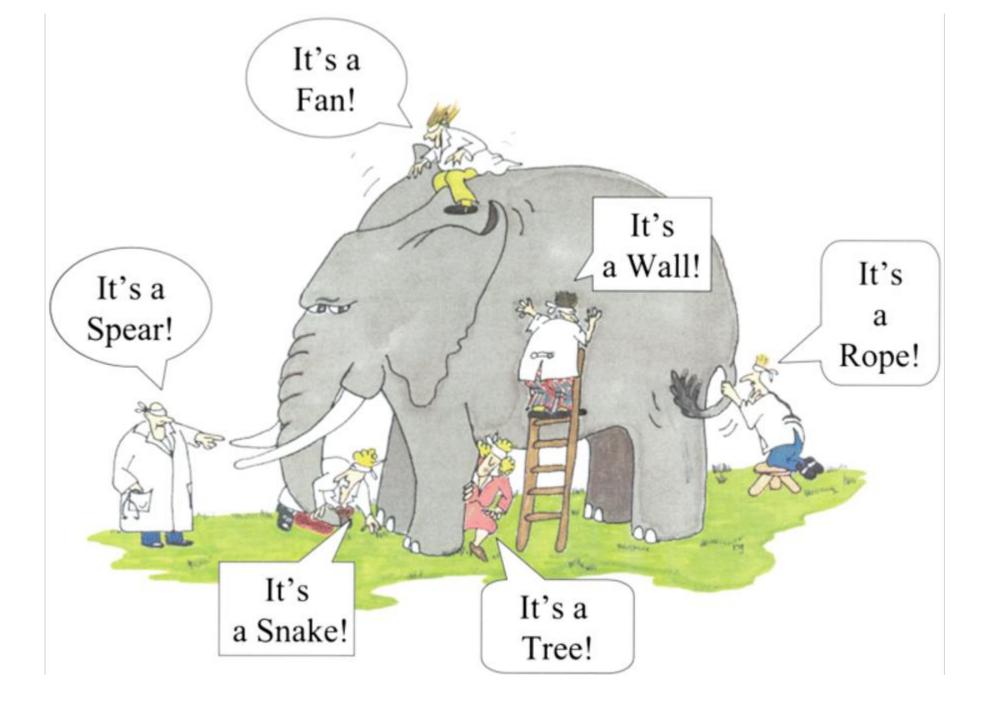


The world's most valuable resource is no longer oil, but data



Agenda

- Definitions ... what do these words mean
- Industrial Hygiene (IH) Data Value Chain ... how does this impact IH
- The Power (and Fury) of Analytics ... what do you need to watch out for
- Toolkits and examples ... what can we do about it



Key Definitions

Big Data

Big data is a term that describes the large volume of data – both structured and unstructured – that inundates a business on a day-to-day basis. But it's not just the amount of data that's important. It is the three Vs – Volume; Velocity and Variety as well as Variability and Complexity that define the term

Machine Learning

A method of data analysis that automates analytical model building. It is a branch of artificial intelligence based on the idea that systems can learn from data, identify patterns and make decisions with minimal human intervention.

Predictive Analytics

Encompasses a variety of statistical techniques from data mining, predictive modelling, and machine learning, that analyze current and historical facts to make predictions about future or otherwise unknown events.

The Role of Data

The 4th Industrial Revolution Started in 2013 (Industry 4.0) – Cyber Physical Systems and takes over from the 3rd Revolution on Computer and Automation

Data waster

Collects data but severely underuse them

Data collector

Collects data but do not consistently maximize their value

Aspiring data manager

Understands value of data and marshals resources to take better advantage

Strategic data manager

Has well-defined data-management strategies that focus on collecting and analyzing the most valuable data









Digital Maturity Model



Digital Maturity Level 1 Moving from Analog

L1 Characteristics

- Isolated Data Sets
- Pen & Paper → Excel
- Bespoke Reports
- Reports On Hard Copies/Hard Drives
- Lack Of Transparency

INCONSISTENT AND INEFFICIENT?



Digital Maturity Level 2 Making the Digital Leap

L2 Characteristics

- Digital Data Collection
- Single Source Of Truth
- Speed Up Decision Making Process (Real Time)

RELIABLE AND

EFFICIENT?

 Consistent And Efficient Reporting

Digital Maturity Level 3 Integrating on Digital

L3 Characteristics

- Connecting Business Data Sources
- Data Insights Leading To Optimization
- Better Visualization (3D) → Stakeholders
- Efficient Decision Making Process

INTEGRATED AND INSIGHTFUL?



Digital Maturity Level 4 Into Data Mining

L4 Characteristics

- Predictive Analytics
- Machine Learning For Data Analysis
- Automated Decision Making

To Artificial Intelligence, Virtual Reality And Beyond

FORWARD-LOOKING AND PROGRESSIVE?

www.erm.com Collection and Interpretation of "Big Data"

Data analytics overview

Descriptive analytics

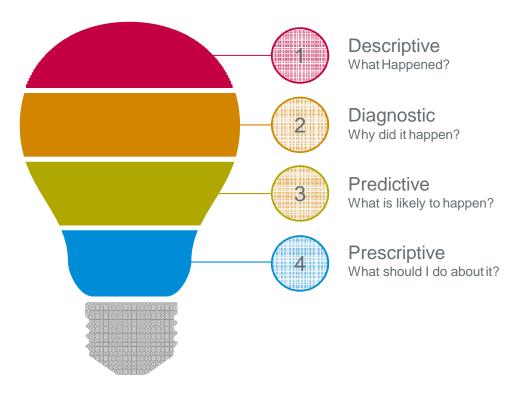
 Describes or summarizes raw data and makes it interpretable

Predictive analytics

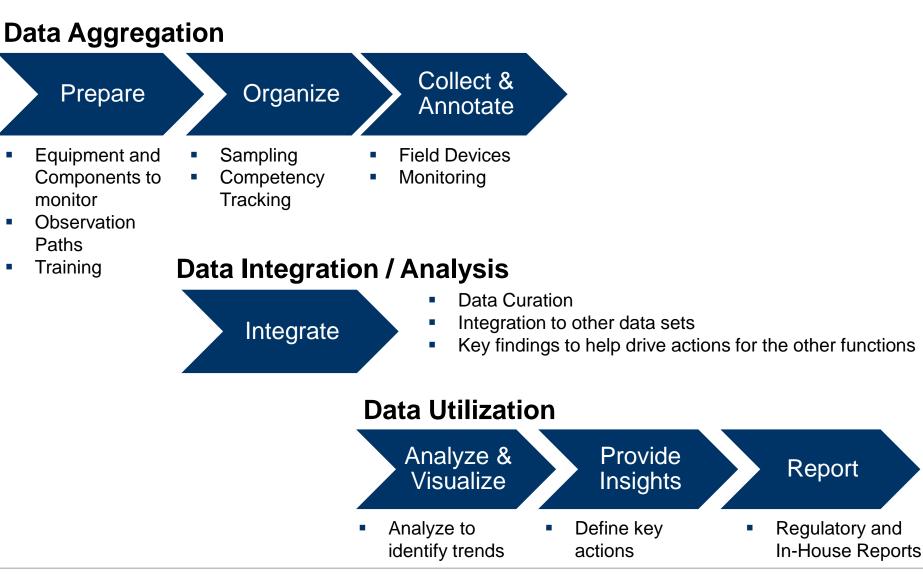
 Helps to understand the future and provide actionable insights

Prescriptive analytics

Quantifies the effect of future decisions



Industrial Hygiene Data Value Chain



Data Aggregation - KISS

Collection and Annotation needs to be the current focus of attention – keeping it simple will be the key

- Utilize sensors currently available
- Leverage what operations is already putting out in place

However a successful and SMART IH program needs to focus on the preparation and organization

- Clear definition of equipment and components to monitor
- Training programs resource identification and planning
- Right to Left Thinking define the "end game"
- Prepare and Organize to ensure that Collect and Annotate is successful

Data Integration / Curation

Avoid data silos

- Standalone systems add to operating complexity
- Unable to fully leverage the value of the data
- Integration can be complex but ways to manage complexity



Cleansing / Curating the data for further analysis:

- Data Profiling Data Quality Safeguards
- Data Cleansing Data Quality Standards and Training

Recognizing "Dirty Data"

- Incorrect Data value is not valid
- **Inaccurate Data** value is not true (still may be valid)
- **Inconsistent Data** same value appears in two places (with minor differences)
- Incomplete Data the data set does not have all the information required for the task or analysis
- Business Rule Violations value is valid and accurate but violates a business rule
- Nonintegrated Data data is not connected appropriately with other data sets which makes analysis and comparisons difficult

Art and Science of Data Curation

Data Profiling

 Uncover your data defects with some data archeology – analyze the data for correctness, completeness, uniqueness, consistency and reasonability

Data Cleansing

 After profiling is complete focus on cleaning the critical and important areas – need to prioritize effort

Data Quality Safeguards

Establish logic, programming, checks to prevent future "dirty data"

Data Quality Standards and Training

• Establish governance process, data stewards and develop training to drive these standards and guidelines across the enterprise

Data Insights

Companies acquire complex and significant amounts of data over the operational lifetime of a site. ERM experience suggests data is rarely stored in a centralised, organised place, can be difficult to locate, is inconsistent, its purpose unclear and the true value of the data in managing risks lost.

ERM's approach is centred on ERM's Digital Foundation a, set of platforms and tools that with our subject matter experts allow us to:



Acquire, transform, and persist data



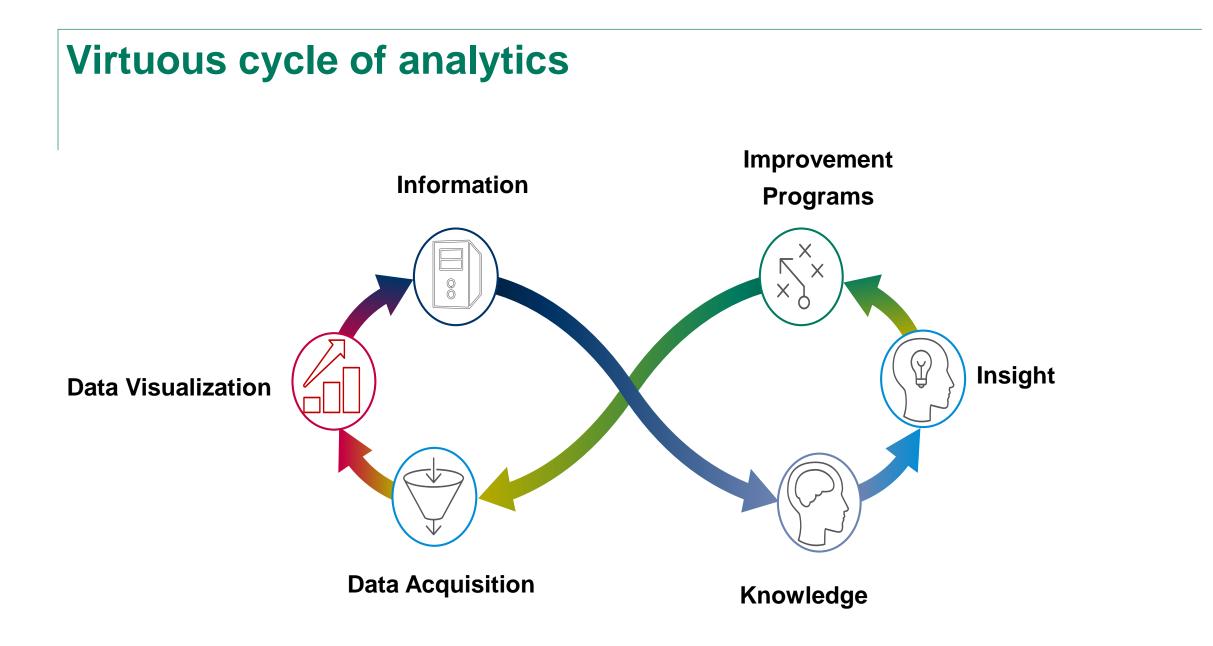
Apply data science techniques to find correlations within data



Apply consultative insight to determine the "so what" in context of the problem that you the client are trying to solve and deliver value



Establish foundation for move to mobile



Data Utilization



Performance Monitoring Dashboard

Audit Samples Dashboard



Power (and Fury) of Analytics

Prediction is very difficult ...

Especially if its' about the future. – Neils Bohr

- *Text message to chemical plant manager*: Chlorine leak expected on line 2 tomorrow. Inspect and repair.
- *High priority email and automatic call to coal mine superintendent*: 83% chance of roof fall on section 4. Evacuate immediately and take corrective actions.
- *Monthly notice to OSHA regional administrator*: HIGH PRIORITY INSPECTION ROSTER: Firms listed below have a greater than 80% probability of violations reflecting hazardous conditions requiring mitigation.

Barriers to implementation:

- Knowledge don't have the data
- Skills don't have the skills / technology to mine the data
- Motivation not keen on acting on the insights
- Inertia fine with current state so not sure the additional effort is worth it
- Privacy GDPR and PII data

Source: Can Predictive Analytics help reduce workplace risk, Dr. Wagner, CDC Blog

Challenges to manage with big data analytics

Causation and Correlation - Post hoc ergo propter hoc

Data Scientists get focused on the statistical analysis of correlation without thoughtful reflection of causation

Sample Size and Data Collection

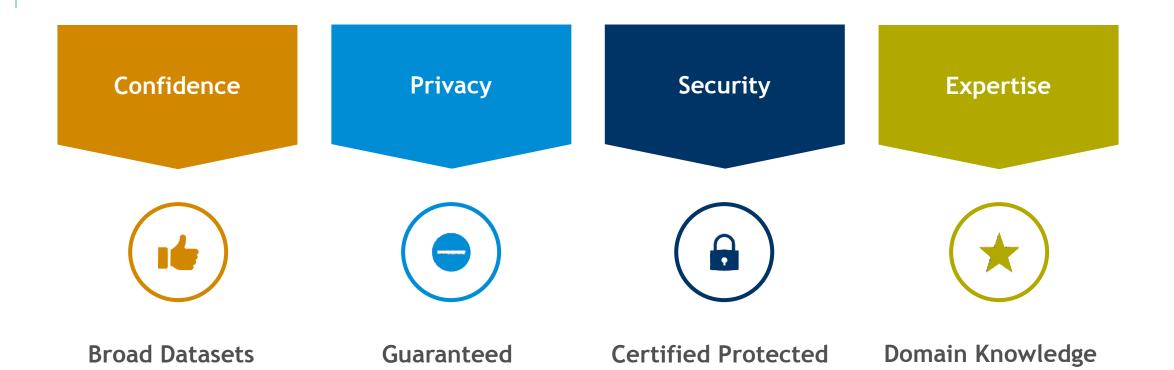
- Need to have a proper and representative sample
- "Stuck in the data forest and lose sight of the smaller but more important trees"
- Issue is starting to become "too much data" rather than "not enough data"

In-built biases in our algorithms

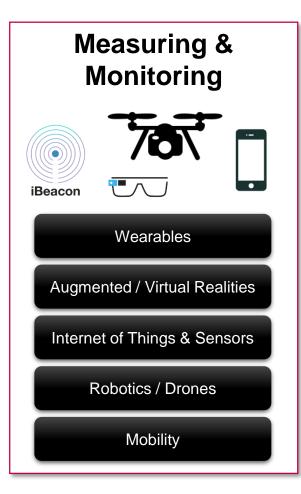
Algorithms are written by human beings – biases do creep in

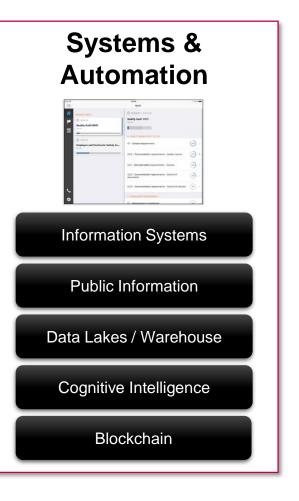
Privacy concerns

Four Pillars of Analytics Excellence



Key digital themes





Reporting, Analytics & Communication Analytics / Data Science **Visualization Tools** Social Media

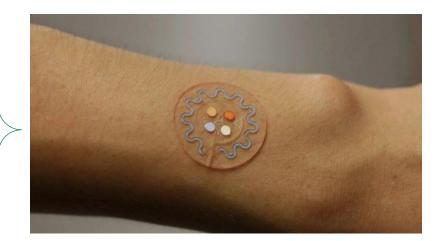
Quick Look: Wearables

Avoiding Dehydration

To avoid dehydration, especially in the Middle East during Ramadan, BHGE is leveraging patches to measure sweat to monitor potential for dehydration. Expanding now to monitor fatigue and stress.

This patch is developed through a consortium of GE, USAF, and universities. A similar patch being developed by a consortium, including Gatorade





Understanding ergonomics

By 2018, 2MM employees will be required to wear health & fitness tracking devices as a condition of employment (per Gartner). Insurance companies, such as Liberty Mutual, are partnering with industry to identify risk areas, such as body movements of workers. Wearables capture movements (e.g., bending, reaching), which are analyzed and then the appropriate equipment and/or work processes are ordered

Quick Look: New Realities

VR Training for Improving Driver Safety

With large exposure of workers on the road, driving as part of their normal job activities, companies, such as UPS and Linde are leveraging VR training to allow employees to become more aware to hazards and risks. This same approach is leveraged highly in the Formula 1 space for impressive results. Research are showing that learnings / time-to-mastery are accelerated by two-thirds.





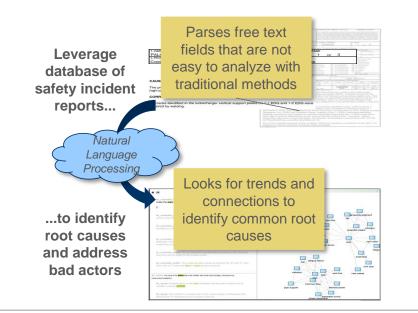
Remote Inspections – BASF

Inspecting and maintaining fire trucks in Germany was expensive - travel expenses, a week of downtime and additional costs. By leveraging AR/VR, the inspections are significantly improved by allowing technician to analyze the truck remotely and identify any parts prior to coming onsite to service the truck

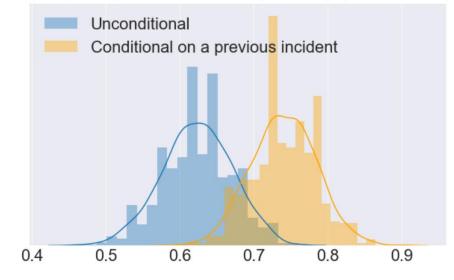
Quick Look: Analytics

Generating Insights

Reviewing existing incident data in combination with a number of different data elements (e.g., weather) and applying data science and subject matter expertise to understand the correlations

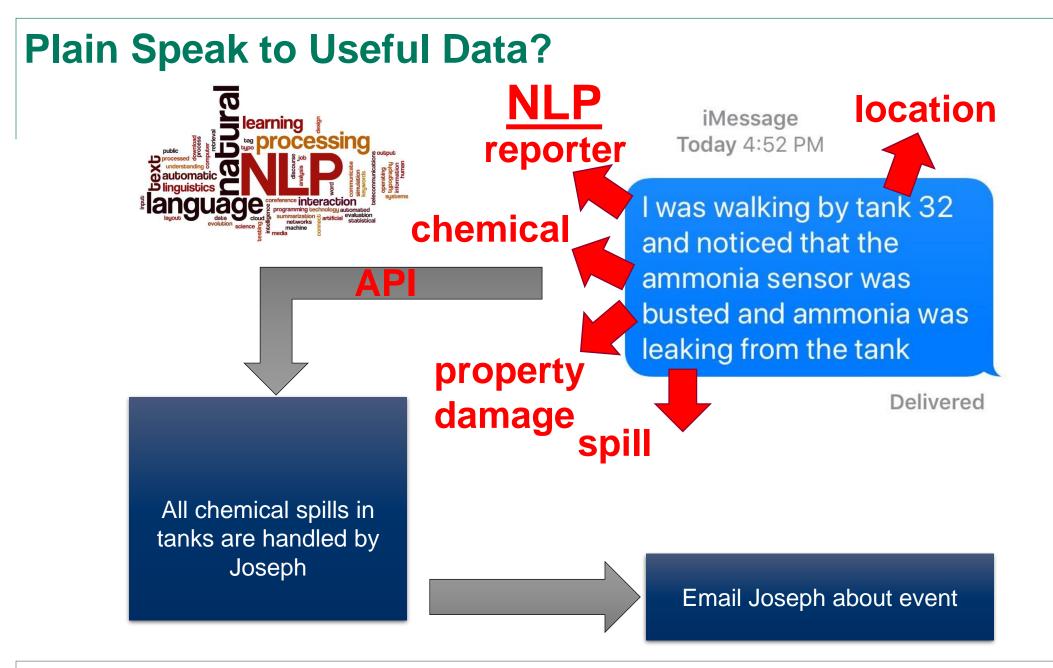


Probability of seeing at least one Actual Incident (serious or above) over the next 30 days



Natural Language Processing

Use of machine learning and natural language processing to improve the ability to analyze large volumes of data and identify hazards, which may be embedded in unstructured free text reports



Summary

- Understand what data is available KISS
- Focus on "Clean Data"
- Define the questions you are trying to answer what is the end-point
- Leverage technology to improve your data collection & analysis
- Be willing to try "fail fast" be AGILE
- Understand how to scale