



### **Data Science:**

Gaining true insights from your data – making the most of your investment

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The business of sustainability

# Health and Safety Moment: Insight through Visualization



 Spill modelling work has identified practical H&S risks for the workforce



### Data Science Introduction



# What We'll Learn About

- 1. The importance of Individual and organisational maturity
- 2. Definition of problem statements/hypotheses and workflows are essential to construct robust/reusable systems
- 3. Don't under-estimate the value in good data and data management
- 4. Big data  $\neq$  Big reports
- 5. Data is a journey

# Moving on from the 1990's





# The Role of Data and the Data Manager

#### Data waster

Collects data but severely underuse them





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 – a Corporate View**



### 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

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# Data + Data Science + SME = Data Insights

Companies acquire complex and significant amounts of data over the operational lifetime of a site. ERM's 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.



- **Collection of Good Data**: Acquire, transform, and persist data appropriately
- ~~~^ \_\_\_\_
- 2. Is it Right? Apply data science techniques to find correlations within data
- 3. What does it tell us? Apply consultative insight to determine the "so what" in context of the problem that you the client are trying to solve and deliver value



# **Role of a Strong Digital Foundation**



### **ERM Digital Foundation**

# **Problem Statements, Hypotheses and Workflow**





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# **Levels of Insight**

**Analytics:** Visualising data – showing trends and indicators – is it performing as it should?

**Machine Learning**: Identifying patterns and probably outcomes – what could happen?

**AI**: Predicting what is going to happen - making decisions on probable outcomes





### Examples: Maturity Stage 2 Reliability & Efficiency Accessing Data to Make Decisions



## Data management

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- Enable mobile ٠ applications
- Improve data access & ٠ consistency



- of risks & liabilities
- Promote collaboration . with consultants
- Improve decision making .



- Allow transition between Consultancies
- Collate legacy data
- Quick response to regulator requests & bespoke regulator views
- Ensure long term access ٠ to project data



# **Case Study: Improving insights and transparency**

**Situation:** Client needed to quickly and efficiently analyze and visualize environmental data.

**Approach:** Develop a centralized repository and visualization tools and create an analysis framework to analyze all environmental data for the organization.

- Developed a framework to analyze environmental data ranging from hourly site-level sensor data to corporate level sustainability metrics; processing 150-200 million data points a year
- Incorporated visualization tools to allow environmental advisors worldwide to quickly trend and analyze performance and understand compliance across all sites, business lines and the corporation
- Developed integration points into control systems to allow operators to react to real-time warning alerts calculated by the system based on sensor data and algorithms



# Automation and data analytics in reporting

### Challenge:

- 200 custom regulatory reports
- Must be completed annually with 4-6 weeks to complete all analyses and reports

### Approach:

Leverage R and Rstudio to create a reporting template that could be automated

### **Benefits:**

- Ability to quickly shift focus to problem sites and manage non-compliance risk.
- Higher confidence and building strong relationship with Regulator
- Overall improvement in the efficiency, accuracy, quality and timeliness of reporting.





Note: The yearly observed mean for lakes during baseline years are represented by symbols only. For lakes during monitored years, the yearly observed mean is shown by symbols, and the mean and 95% confidence interval estimated by model fitting is represented by curved horizontal lines and vertical bars respectively.

0.0.0.7 Minimum Detectable Differences

2.1.2 Nutrients

2.1.3 Metals

The estimated minimum detectable difference in mean ph for each monitored lake in 2017. Reference lakes are shown for comparison.



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### Examples: Maturity Stage 3 Data Insights Identifying Patterns and Reacting



### **Data accessibility**

Decomposition of additive time series 140 observed randomseasonal trend observed mm homm 8 8 2020 8 randomseasonal trend 8 8 8 15 30 문 <del>5</del> 0 ę 20 -15 5  $\sim$ AA 1147 . 무 - 100 Seasonal plot: Asia Pacific Seasonal plot: North\_America <del>6</del> year year 75 -2011 - 2011 100 -- 2012 2012 2013 2013 - 2014 - 2014 50 -- 2015 2015 2016 2016 50 2017 2017 25 -NA APAC Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec Month Month Statistical Analysis (Snapshot)

Decomposition of additive time series





Converting to a Visual

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# **Dashboards for data presentation & corporate overviews**



# **Dashboards for managing operational risk**





# Examples: Maturity Stage 4 Data Mining Creating Predictions from Data



# **Data Insights**

- How do we get true value out of data?
- Developing a good data workflow can enable us to:



Acquire, transform, and store data



Apply data science techniques to find correlations and assess the strength of relationships



Identify the different combinations of factors driving outcomes



Detect and understand the early warnings and weak signals which point to a vulnerability, an emerging trend or a hard to predict event



Provide in-depth assessment by applying consultative insight to determine the "so what" in context of the problem that the client is trying to solve and deliver value

# Data insight approach



# Data + Data Science + SMEs = Insights



The targeted improvement roadmap developed as a result of this approach focuses on one or more of three primary areas of concern:

- Organisational change Identifying if the right data is being acquired and effectively delivered to the right areas of the organisation to affect timely decision support.
- Functional change Identifying where technology and data can be more effectively utilised to improve the internal service delivery of specific functions across Health, Safety and Environment.
- Systems change Identifying where system improvements could be implemented to drive out complexity and cost and/or improve performance.



# **Case Studies**

The two case examples that follow come from one of our international mining clients. The first looking at the organisation's regional operation, the second focusing on a single operation.

These outputs were two of 20 key insights provided by ERM in a work programme delivered over 14 weeks for the Group level HSE Director.

### Case study High potential risk management

Following one HiPo event, the data revealed that likelihood of a second event happening within 30 days was greater.

### Findings

On a "typical" day (blue curve) there is a probability of about 62% that a severe incident (serious or above) will happen in the next 30 days.

However, If an incident has just happened (orange data), this probability rises to about 74%.

### Value

Our analysis encouraged the organisation to revisit the processes they put in place in their global mining operations immediately following a HiPo event.



### Case study Production impacts on H&S performance

High level analysis of key production metrics in one asset demonstrated an inverse relationship between production and incident rates.



Deeper analysis investigating the following potential contributory and causal factors was also considered:

- Type of work activities
- Staff turnover rates
- Ratio of contract and new hires in workforce
- Weather/environmental conditions
- Impact of H&S campaigns.



# Further Advances: Visual Communication



# **Better reports and communicating data**





# Thank you

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