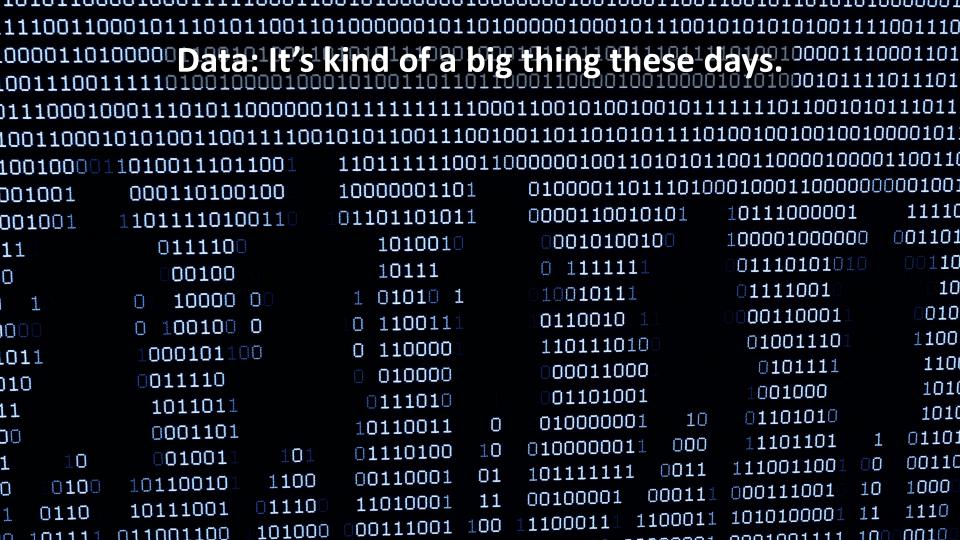
# Beyond Statistical Literacy: Open Data for Teaching, Collaboration, and Profit

#### Mike Smit

School of Information Management Faculty of Management







## Transition from data scarcity to data plenty

Scarcity

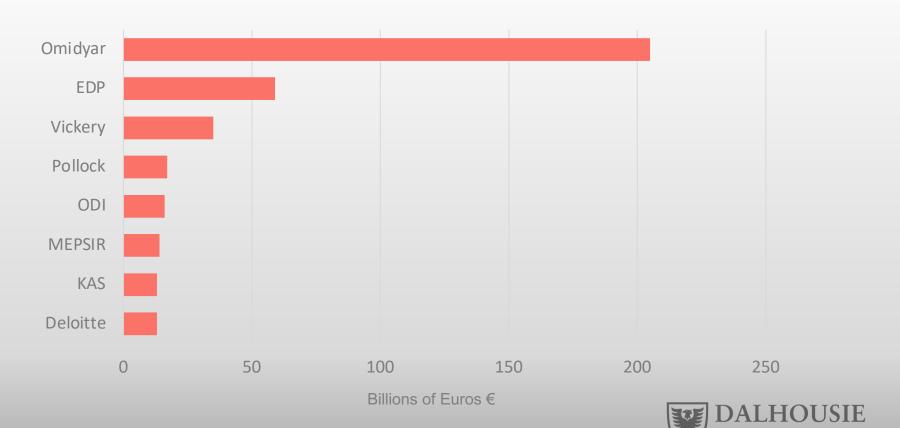
Plenty

- Data is an asset & must be protected
- Challenge is accessing the data you need
- Mostly competition
- Short-term gain
- Key issue: data hoarding

- Relationships are an asset, data is a networking tool
- Challenge is finding the data you need
- Mostly cooperation
- Long-term gain
- Key issue: data waste



## Various estimates for the value of open data to EU



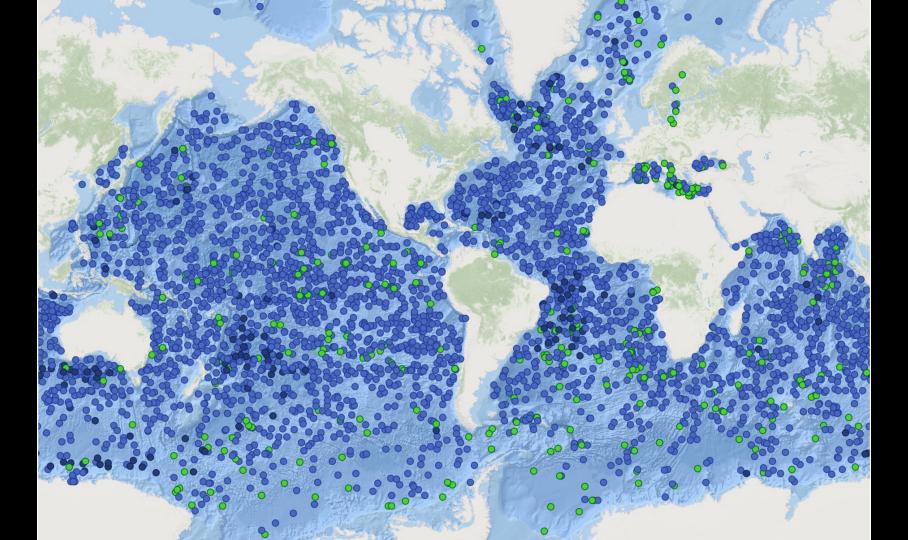
Compiled by: European Data Portal. (2017). Analytical Report 9: The Economic Benefits of Open Data.

# Schrödinger's Data: open data is worthless and priceless at the same time









# So how do we equip society to unlock this value?

• Some thoughts, based on five main categories of information



# Strategies and Best Practices for Data Literacy Education

- SSHRC Knowledge Synthesis
- The transdisciplinary Team:

James Rothwell	Hossam Ali-Hassan
MLIS/MPA graduate student	Business (MIS)
Dean Irvine	Daniel Kelley
English	Oceanography
Mike Smit	Brad Wuetherick
Information Management	Centre for Learning & Teaching
	MLIS/MPA graduate student  Dean Irvine English  Mike Smit





# Strategies and Best Practices for Data Literacy Education

**Knowledge Synthesis Report** 

Chantel Ridsdale, James Rothwell, Mike Smit, Hossam Ali-Hassan, Michael Bliemel, Dean Irvine, Daniel Kelley, Stan Matwin, and Brad Wuetherick



# **Databilities**®

A data literacy competency framework



# [PDF] Strategies and best practices for data literacy education: Knowledge synthesis report

C Ridsdale, J Rothwell, M Smit, H Ali-Hassan... - 2015 - dalspace.library.dal.ca

BACKGROUND: We begin with a definition, synthesized from existing literature and refined based on expert input: **Data literacy** is the ability to collect, manage, evaluate, and apply **data**, in a critical manner. It is an essential ability required in the global knowledge---based ...

☆ ସସ Cited by 121 Related articles All 6 versions >>>

**Chantel Ridsdale, Adrienne Colborne, and Michael Smit Dalhousie University** 

Proficient Use of Open Data Requires These Core Information Skills: An Open Data Community Perspective

#### Abstract:

Expanding access to open data, such as government data and research data, requires that we consider how citizens and stakeholders can best access the value these data hold. Should individuals rely on an intermediary to create information products from the data, or should they dive in and work with raw data? Building on previous work defining a core set of data literacy skills, we convened a workshop with 34

open data professionals to define the core set of skills for working with open data: "open data literacy".

Analysis of their perspectives reveals a focus on non-technical skills, like creativity, curiosity, and critical thinking as a priority over technical skills like coding and visualization. We describe their perspective in





#### Bachelor of Management

MGMT 2605 Working with Data Fall 2020

#### RECOGNITION OF MI'KMAQ TERRITORY

Dalhousie University is located in Mi'kma'ki, the ancestral and unceded territory of the Mi'kmaq. We are all Treaty people.

The Elders in Residence program provides students with access to First Nations Elders for guidance, counsel and support. Visit the office in the Indigenous Student Centre, 1321 Edward Street, or email Elders@dal.ca.

Instructor: Colin Conrad Pronouns: he/him

Office: Kenneth C. Rowe Management Building, 4020

Telephone: (902) 494-8378 E-mail colin.conrad@dal.ca

#### **About Your Instructor:**



Colin Conrad, Ph.D., Assistant Professor School of Information Management, Dalhousie University dal.ca/execed/big-data

# DATA LITERACY & BUSINESS ANALYTICS WORKSHOP



data literacy: The ability to create, comprehend, and communicate data.

The ability to collect, manage, evaluate, and apply data, in a critical manner.



"By 2018, the United States alone could face a shortage of 140,000 to 190,000 people with deep analytical skills..."

McKinsey Global Institute



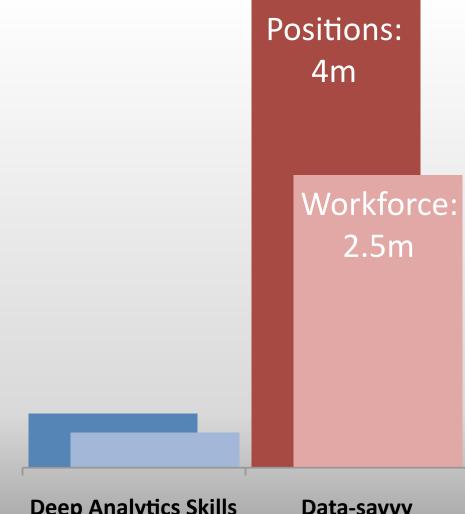
# Positions: 465k

Workforce: 300k

# **Skills Gap**

 Predicted for US in 2018 by McKinsey Global Institute







### Closing Canada's **Big Data Talent Gap**



**Canada's Big Data Consortium** October 2015

We would like to thank the following organizations for their leadership and in-kind contributions to Canada's Big Data Consortium, Canada's Big Data Talent Gap Study, and to this paper, "Closing Canada's Big Data Talent Gap."

























# 21<sup>st</sup> Century Literacies

- Data Literacy can be understood as part of a broader suite of literacies
- Information literacy
- Technological/digital literacy
- Statistical/numerical literacy
- Visual literacy
- Scientific literacy
- Cultural literacy
- Metaliteracy/transliteracy



## Data Evaluation

#### Data Tools

- Management •Basic Data Analysis
  - Data Interpretation (Understanding Data)
  - •Identifying Problems Using Data
  - Data Visualization
  - Presenting Data (Verbally)
  - Data Driven
     Decisions Making
     (Making decisions
     based on data)

- Data Application
- Critical Thinking
- Data CultureData Ethics
- Data Citation
- Data Sharing
- Evaluating Decisions Based on Data

#### Data Collection

Data Discovery and Collection

Conceptual

Framework

- •Evaluating and Ensuring Quality of Data and Sources
- Data ManipulationData Conversion

Data

(from format to format)

Data Organization

- •Metadata Creation and Use
- Data Curation,
   Security, and Re-Use
- Data Preservation

Conceptual Competencies

**Core Competencies Advanced Competencies** 



## **Conceptual Framework**

Introduction to Data

- Knowledge and understanding of data
- Knowledge and understanding of the uses and applications of data

**Conceptual Competencies** 

**Core Competencies Advanced Competencies** 

## **Data Collection**

**Data Discovery** and Collection

- Performs data exploration
- Identifies useful data
- Collects data

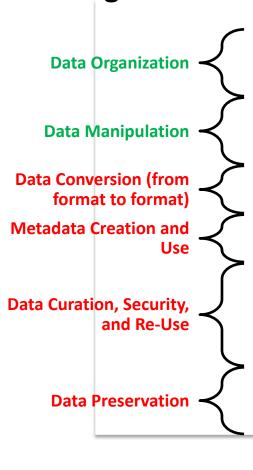
Evaluating and Ensuring Quality of Data and Sources

- Critically assesses sources of data for trustworthiness
- Critically evaluates quality of datasets for errors or problems

**Conceptual Competencies** 

Core Competencies
Advanced Competencies

**Data Management** 

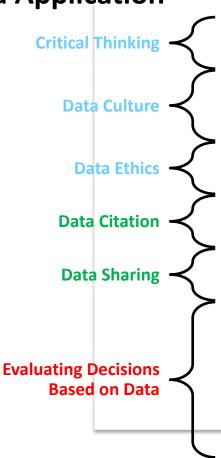


- Knowledge of basic data organization methods and tools
- Assesses data organization requirements
- Organizes data
- Assesses methods to clean data
- Identifies outliers and anomalies
- Cleans data
- Knowledge of different data types and conversion methods
- Converts data from one format or file type to another
- Creates metadata descriptors
- Assigns appropriate metadata descriptors to original data sets
- Assesses data curation requirements (e.g. retention schedule, storage, accessibility, sharing requirements, etc.)
- Assess data security requirements (e.g. restricted access, protected drives, etc.)
- Curates data
- Assesses requirements for preservation
- Assesses methods and tools for data preservation
- Preserves data

Conceptual Competencies
Core Competencies
Advanced Competencies

#### **Conceptual Competencies Data Evaluation Core Competencies Advanced Competencies** • Knowledge of data analysis tools and techniques Knowledge/Tasks **Data Tools** •Selects appropriate data analysis tool or technique • Applies data analysis tools and techniques Develops analysis plans Applies analysis methods and tools **Basic Data Analysis** Conducts exploratory analysis Evaluates results of analysis •Compares results of analysis with other findings •Reads and understands charts, tables, and graphs **Data Interpretation** •Identifies key take-away points, and integrates this with other important information (Understanding Data) •Identifies discrepancies within the data •Uses data to identify problems in practical situations (e.g. workplace efficiency) **Identifying Problems** •Uses data to identify higher level problems (e.g. policy, environment, scientific experimentation, **Using Data** marketing, economics, etc.) •Creates meaningful tables to organize and visually present data •Creates meaningful graphical representations of data **Data Visualization** • Evaluates effectiveness of graphical representations Critically assesses graphical representations for accuracy and misrepresentation of data •Assesses the desired outcome(s) for presenting the data Assesses audience needs and familiarity with subject(s) **Presenting Data** • Plans the appropriate meeting or presentation type (Verbally) •Utilizes meaningful tables and visualizations to communicate data Presents arguments and/or outcomes clearly and coherently • Prioritizes information garnered from data **Data Driven Decisions** Converts data into actionable information Making (Making • Weighs the merit and impacts of possible solutions/decisions decisions based on data) • Implements decisions/solutions

**Data Application** 



- Aware of high level issues and challenges associated with data
- Thinks critically when working with data
- Recognizes the importance of data
- Supports an environment that fosters critical use of data for learning, research, and decision-making
- Aware of legal and ethical issues associated with data
- Applies and works with data in an ethical manner
- Knowledge of widely-accepted data citation methods
- Creates correct citations for secondary data sets
- Assesses methods and platforms for sharing data
- Shares data legally, and ethically
- Collects follow-up data to assess effectiveness of decisions or solutions based upon data
- Conducts analysis of follow-up data
- Compares results of analysis with other findings
- Evaluates decisions or solutions based on data
- Retains original conclusions or decisions, or implements new decisions/solutions Conceptual Competencies

**Core Competencies Advanced Competencies** 

Skill	Votes
Basic statistics & math	19
Pattern Recognition	17
Basic computer literacy	16
Ability to judge sources; where is the data coming from	15
Skepticism	13
Communication Skills	13
Data Visualization	11
Patience & Focus	10
Curiosity	10
Awareness of bias	10
The top individual skills required to work directly with open data, as identified by a panel of experts.	

Skill Area	Number of Skills	Total Votes
Critical Thinking	5	47
Numeracy	5	41
Awareness	7	38
Analysis	7	35
Communication	5	34
Attitude	7	32
Computer Knowledge	6	27
TOTAL	42	254



# How do we produce more data-literate graduates?





- We have data literacy expertise
- We teach elements of data literacy every day
- Many graduates are data literate without knowing it!



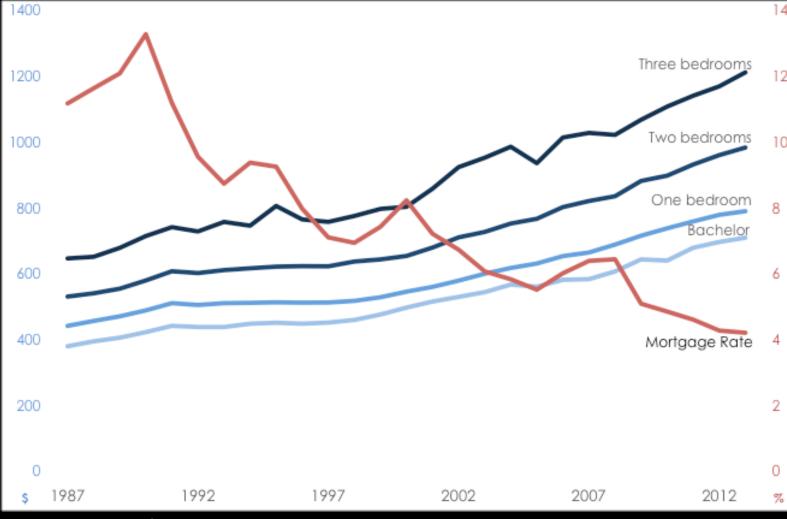
- No systematic, crossdiscipline best practices
- No universal standard
- Not always taught as transferable skill



# 1. Project-based learning with some choice

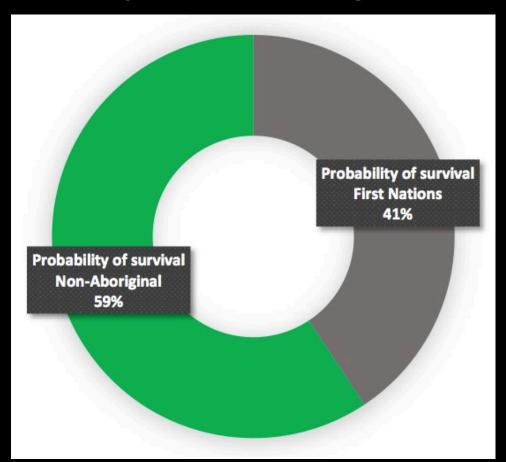
- Projects that include a wide range of investigation and have real-world applicability will solidify the connection between process/theory and practice. The project will allow evaluators the chance to assess skills practically, instead of formally.
- Projects should offer students the opportunity to go further than you expect.
- We love to use open data for this!



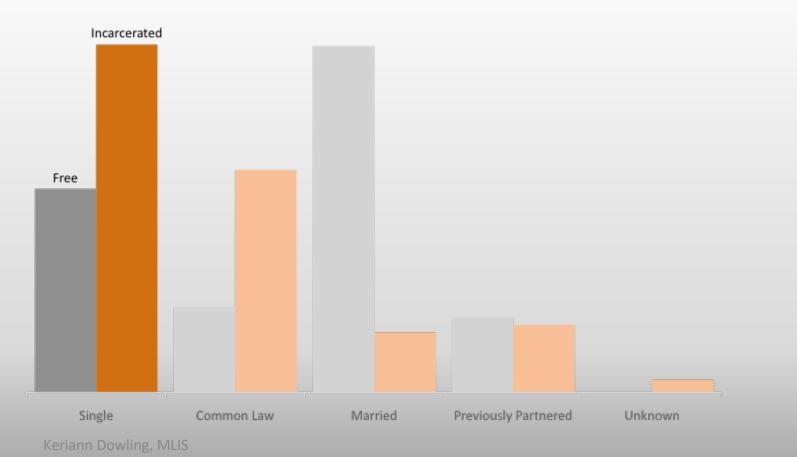


Carlisle Kent, MLIS /MREM 2016

## **Probability of Women Surviving 80 Years**



Canada's Most Eligible Bachelors:
Marital Status for Incarcerated Men vs. Canadian Men Over 18



# 2. Link training to things that matter to students

- The benefits of data, and data skills, must be clearly stated from the beginning. This is particularly true for mid-career learners, who will be more willing to invest their limited time and effort if they see the opportunity to help their community, industry, family, or others.
- The motivation should be intrinsic for students





# VIRTUAL REALITY OCEAN CHALLENGE

JANUARY 17 & FEBRUARY 6
SHIFTKEY LABS





# 3. Modular, just-in-time learning (as possible)

- The "learn this now because you'll need it later" has always been a stretch, and now even more so
- Making training available in the moment when it is needed for things that matter intrinsically to students
- Integrated into other courses in a program / curriculum
- Related to the concept of learning / tutoring centres and similar



# 4. Hands-on learning, and make mistakes okay

 Hands-on learning in workshops and labs provides students with the necessary practical experience needed to fully understand a technical skill; students need the chance to figure out processes and methods on their own and make mistakes to readjust their own understanding. Mechanics are very important in data literacy; practice is required. Making mistakes can be frustrating, but will encourage critical thinking and problem solving.



# 5. Laddering / iterative learning

 Allows students to build upon previously learned skills, encouraging process over memorization or following rigid instructions, and ultimately making learning an unfamiliar concept more manageable. Beginning small and working up to the more complicated tasks allows students to have confidence in their abilities.



# When?

• Before post-secondary?







# **System Barriers to Effective Data Literacy Instruction**

- Lack of support/conditions for innovative learning, e.g. creativity, risk-taking (Johnson, Adams Becker, Estrada, & Freeman, 2015).
- Lack of university faculty trust in benefits of data literacy (Ikemoto & Marsh, 2008; Johnson, et al 2015).
- Mixed support for different data literacy competencies, e.g. teaching some aspects of data literacy, but not others (Carlson, Johnston, Westra, & Nichols, 2013)
- Abstract nature and perceived complexity of data literacy and related concepts (Qin & D'Ignazio, 2010; Twidale, Blake, & Grant, 2013).
- Misconception that the Net Generation/Digital Natives are inherently more knowledgeable technically than past students (Thompson, 2012).

# Operational Barriers to Effective Data Literacy Instruction

- Modifying existing curriculum to include room for course(s) on data literacy (Hunt, 2004; Teal, Cranston, Lapp, White, Wilson, Ram, & Pawik, 2015)
- Not enough skilled professors, or not enough time to learn the required skills to effectively teach a data literacy course (Boyles, 2012; Carlson, et al, 2013; Teal, et al, 2015)
- Poor communication regarding student learning goals, as well as a lack of communication/collaboration between professors relating to what has been covered in courses, and what should be developed for future courses (Cowan, Alencar, & McGarry, 2014; Teal, et al., 2015)
- Differing levels of comfort and experience with technology and/or data, differing levels of education (graduate vs. undergraduate), and age/generational differences between students (Jones, Ramanau, Cross, & Healing, 2009; Shorish, 2015)



# **Questions / Discussion Welcome!**

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- @michael\_smit



# References

 Full references are available in the KS report at http://dalspace.library.dal.ca/handle/10222/64578

