# FIT5145 Introduction to Data Science Module 2 Data Models in Organisations 2019 Lecture 3

Monash University

#### **Discussion: Motion Charts**

Motivation: <u>TED talk by Hans Rosling</u>

#### Advantages:

- time dimension allows deeper insights & observing trends
- good for exploratory work
- motion allows identification for this out of common "rhythm"
- "appeal to the brain at a more instinctual intuitive level"

#### Disadvantages:

- not suited for static media
- display can be overwhelming, and controls are complex
- not suited for representing all types of data, e.g. other graphics might be suitable for business data
- "data scientists who branch into visualization must be aware of the limitations of uses"

#### Discussion: Data Science Jobs

#### Data Science Job Market in Australia

- smaller (per capita) market compared to USA & UK, where giant industry players are making better use of Data Science
- currently lacks proper synchronization between academia, practice and policy

#### Job Adverts

- communication skills and domain expertise are rated highly different jobs require different toolset skills
- see Adzuna's CV upload page for an interesting application!

#### **FLUX Question**

Data scientists are primarily people who develop insights with data.

A. TRUE

B. FALSE

#### Unit Schedule: Modules

Module	Week	Content		
1.	1	overview and look at projects		
	2	(job) roles, and the impact		
2.	3	data business models		
	4	application areas and case studies		
3.	3. 5 characterising data and "big" da			
	6	data sources and case studies		
4.	7	resources and standards		
	8	resources case studies		
5.	9	data analysis theory		
	10	data analysis process		
6.	11	issues in data management		
	12	GUEST SPEAKER & EXAM INFO		

### Learning Outcomes (Week 3)

By the end of this week you should be able to:

- Explain general models for understanding businesses and decision making:
  - Explain values chain and our standard model for data science process
  - ► Identify different analytic levels in a data science project and comprehend the difference between levels
  - Learn how Influence Diagram can be used as a tool to model decision making
- Analyse business models specific to data science in different organisations
- Comprehend more sophisticated group-by operations and graphing in Python

#### Case Study: City Science

#### City X

- ► is a research lab at UNSW, formerly at Monash
- focuses on improving scientific understanding of cities and providing new insights into developing a data-driven approach to design, plan, and operate future cities
- ► see video on *Pedestrian traffic* in Melbourne

#### Case Study: Is Big Data Better?

#### See this Strata-Hadoop video from 2013:

- <u>"Is Bigger Really Better?</u> Predictive Analytics with Fine-grained Behavior Data"
- by <u>Foster Provost</u>, author of the book "Data Science for Business"

## Data and Decision Models (ePub section 2.1)

## some general models for understanding businesses and decision making:

- ► value chains:
  - sequence of processes done to create value
- analytic levels:
  - to broadly classify kinds of analysis
- influence diagrams:
  - method for modelling decision making
  - will extend later to model learning

## Data and Decision Models: value chains

sequence or processes for data science

#### Our Standard Value Chain

Collection: getting the data

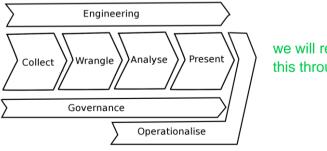
Engineering: storage and computational resources

Governance: overall management of data Wrangling: data preprocessing, cleaning

Analysis: discovery (learning, visualisation, etc.)

Presentation: arguing that results are significant and useful

Operationalisation: putting the results to work



we will refer to this throughout!

## Data and Decision Models: Analytic levels

descriptive terms used to broadly classify kinds of analysis

### **Analytic Levels**

- Descriptive Analytics: gain insight from historical data
  - plot sales results by region and product category
  - correlate with advertising revenue per region
- Predictive analytics: make prediction using statistical and machine learning techniques
  - predict next quarter's sales results using economic projections and advertising targets
- Prescriptive analytics: recommend decisions using optimization, simulation, etc.
  - recommend which regions to advertise in given a fixed budget

primarily a descriptive classification for general discussions

#### **FLUX Question**

Which of the following is an optimization or prescriptive analytics task (as opposed to a predictive analytics task)?

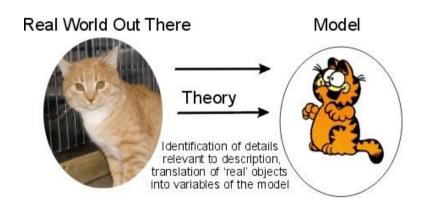
- A. Recommending a traffic route based on prior data for the time of data and incident reports.
- B. Predicting travel time of multiple traffic routes

## Data and Decision Models: Influence diagrams

#### method for modelling data and decision making

- what do we know?
- ► what don't we know?
- what value can we get?
- ► what decisions do we have to make?

### Modelling

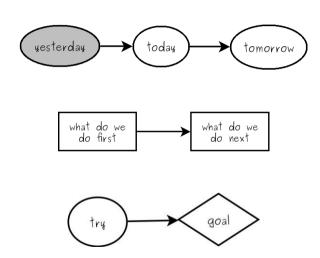


from the BackReaction blog by Sabine Hossenfelder

### Modelling

- "all models are wrong, but some are useful"
  - ... George Box
- "the approximate nature of the model must always be borne in mind"
  - ... George Box
- "the purpose of models is not to fit the data but to sharpen the questions"
  - ... Samuel Karlin

### Influence Diagrams



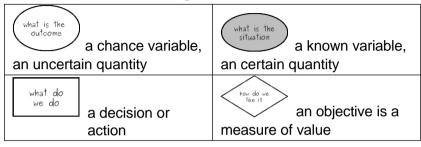
### **Motivating Influence Diagrams**

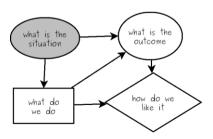
Influence Diagrams (a.k.a Decision Graphs) are

- directed graphical model with 4 types of nodes:
  - chance nodes, known variable nodes, action/decision nodes and objective/utility nodes
- model the "influences", "causes", random ("chance") outcomes, "actions", "goals" involved in a decision problem
- provide a coarse abstraction, a conceptual model

a conceptualisation aid to get you thinking about actions, values, and unknowns

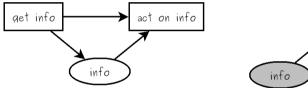
#### Influence Diagrams

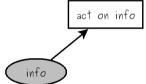




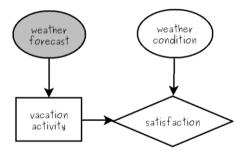
nodes are strung together in a directed graph to convey influence

### Influence Diagrams – Arcs

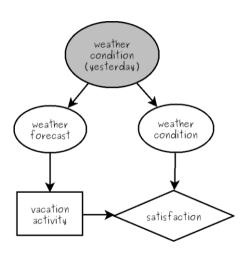




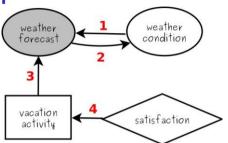
#### **Last Minute Vacation**



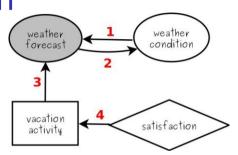
### Last Minute Vacation (cont)



Bad Arcs for Last Minute Vacation

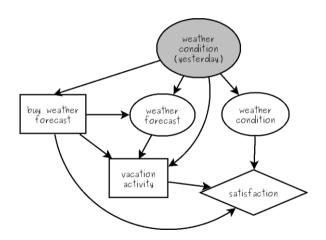


## Bad Arcs for Last Minute Vacation



- Weather cannot cause its forecast!
- The forecast cannot cause the weather!
- You're decision to go on vacation follows in time after you have obtained forecast.
- The success (failure) of the vacation follows in time after your decision.

## Last Minute Vacation with Forecast



## Node Types

chance variable	known variable	decision or action	objective
what is the outcome	what is the situation	what do we do	kow do we like it

When do we connect an arc to a node?

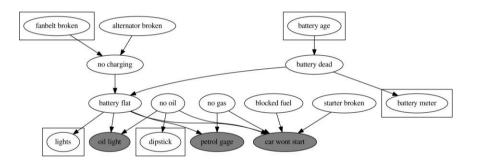
Chance variable: connect node A to chance node B if changes to the value of A can "cause" changes in B;

Known variable: same as chance node

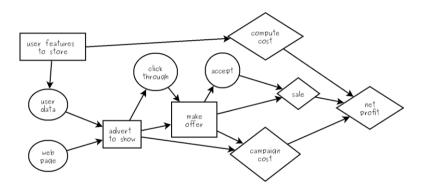
Decision: connect node A to decision node B, if variable A is used when making decision B;

Objectivity: connect node A to objective node B if variable A is used when evaluating the value of the objective (e.g. quality or cost)

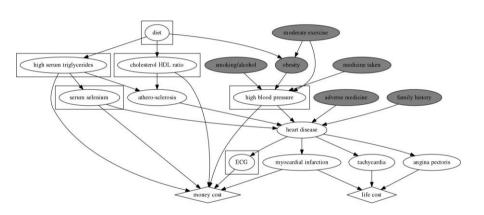
#### Your Car Wont Start



### **Internet Advertising**



#### **Heart Disease**



#### **FLUX Question**

#### An Influence Diagram:



- A. is a model giving possible situations or outcomes.
- B. consists of nodes and arcs.
- C. is an alternative to decision tree.
- D. consists of nodes and arcs and is an alternative to decision tree.

## Business Models with Data (ePub section 2.3)

what kinds of businesses do we have operating in the Data Science world?

#### **Business Models**

#### From Wikipedia:

A <u>business model</u> describes the rationale of how an organization creates, delivers, and captures value, in economic, social, cultural or other contexts.

#### Examples of general classes:

- retailer versus wholesaler
- luxury consumer products
- software vendor
- service provider

What kinds of businesses do we have operating in the Data Science world?

## Business Models for Data Science

Many Data Science companies fit into traditional IT business models.

- software as a service (SaaS)
- consulting
- customer relationship management
- etc.

**e.g.** SAS is both a software vendor and a consultancy, both traditional IT business models

But there are business models somewhat unique to data-based businesses like data science.

## Business Models with Data: Data business models

what are some business models specific to data science?

### **Bloomberg Terminal**



by Jm3 CC BY-SA 3.0, via Wikimedia Commons

# **Bloomberg Terminal (cont)**

#### the Bloomberg Terminal:

- a computer system provided by Bloomberg L.P
- enables professionals to monitor and analyze real-time financial market data
- also place trades on the electronic trading platform
- is a proprietary secure network

#### Questions:

- where does the data originally come from?
- why don't users of the terminals get their data from the original source?
- why wouldn't people who sell the data to Bloomberg set up a similar service themselves?

# Bloomberg Terminal (cont)

Bloomberg provides an information brokering service.

broker ::= a person who buys and sells goods or assets for others

#### Amazon.com

Intro. to Data Science, (c) Warans Braintine, 1804 600 2018-way



Toshiba Satellite L655-

## Amazon.com



## Amazon.com (cont.)



- an assembly line for the retail industry, with support for embedded online retailers
- ► huge stock of books, DVDs, CDs, etc. easily searchable
- extensive customer reviews

## Amazon.com (cont.)

Information-based differentiation: satisfies customers by providing a differentiated service:

- superior information including reviews about products
- superior range

Information-based delivery network: they deliver information for others; retailers in the Amazon marketplace get:

- customers directed to them
- other retailers' support

### LexisNexis

See <u>LexisNexis</u>, provides world's largest electronic database for legal and public-records related information.

Information provider: business selling the data it collects

- ► like a traditional business model, selling data not widgets
- fasting growing segment of the IT industry post 2000 (cited by Evan Quinn's blog post on Infochimps. com April 2013 "Is Big Data the Tail Wagging the Data Economy Dog?", now offline)
- some call this the data economy
- e.g <u>data brokers sell consumer data</u> to major retailers or internet companies

### **Data Business Models**

- information brokering service: buys and sells data/information for others.
- information-based differentiation: satisfies customers by providing a differentiated service built on the data/information.
- information-based delivery network: deliver data/information for others.
- information provider: business selling the data/information it collects.

<u>"What a Big-Data Business Model Looks Like"</u> by Ray Wang in the Harvard Business Review claims these are unique in the data world.

## Business Models with Data: Intelligent systems business models

more generally, what are some intelligent systems business models?

## Intelligent Systems Bus. Models

From <u>Machine Intelligence in the real world</u> by Shivon Zilos, 26th Nov. 2015.

#### Business Models used by Intelligent Systems companies:

Data providers collect dataset (e.g. satellite imagery, vineyard irrigation)

Alchemists promise to turn your data into gold (e.g. self-service APIs)

Gateways create new use cases from specific data types (e.g. image, audio, video, genomic data)

Magic Wands fix a workflow using software as a service (SaaS) tools (e.g. help recruiters write better job descriptions)

Navigators autonomous systems for the physical world (e.g. self-driving cars)

Agents create cyborgs and bots to help with virtual tasks (e.g. customer service realtime chat)

# A brief Introduction to Python for Data Science

## **FLUX Question**

#### What is .ipynb?



- A. An illegal file extension.
- B. Interactive Python NoteBook.
- C. Intelligent Python Nota Bene.
- D. Typo, it should be 'pinyin'

## **FLUX Question**

#### What is a dataframe?



- A. An array.
- B. A list.
- C. A theory about data.
- D. A structure that stores tabular data

# Next Week: Applications Areas (ePub section 2.5)

#### Homework:

- read <u>"The Mayor's Geek Squad"</u> from New York Times
- view <u>"How People Spend Their Time Online"</u> by GO-Gulf (infographic on a blog)
- read <u>"Data Science and its Relationship to Big Data</u> and Data-Driven Decision Making," by Foster Provost and Tom Fawcett. *Big Data*. March 2013, 1(1): 51-59 (8 page PDF).