

Take a look at the stages of implementation for data and analytics.

- ➤ What stage is your organization in?
- ➤ How would you get to an 'at-scale' implementation?

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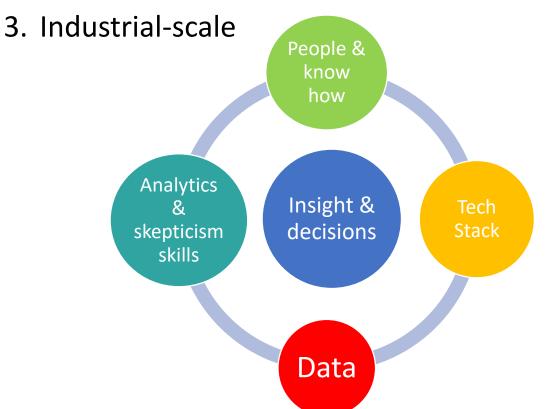
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#### Outline

How do organisations implement data and analytics at various levels of ambition

- 1. Pilot
- 2. Mid-scale



Beyond the learnings we gain as we evolve from one stage to the next, use of data science in a policy setting requires

- Scale
- Trust
- Stakeholder diversity
- Reliability
- Built to evolve



#### Pilots – a bit of luck



idea somewhere





Yields an outcome that impresses



## Pilots – the 4 pillars are expressed, but not reproducible

People & know how

One Business leader
One Data Scientist
(Often some external spark is involved)

**Tech Stack** 

Ranges from available in-house software to public sources

**Data** 

Often a one-time, eye watering effort

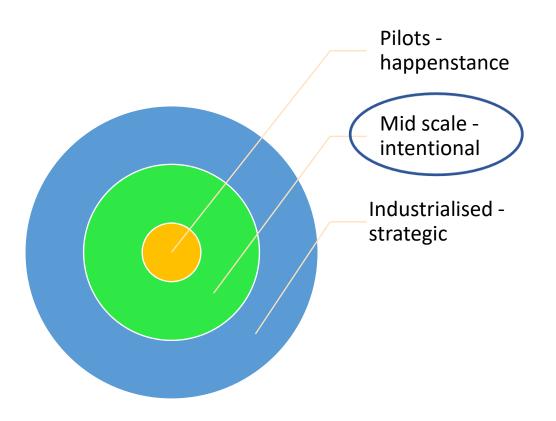
Analytics & skepticism skills

Point solution
Validation, reporting and archiving at the discretion
of the data scientist

In other words, it's the outcome that counts

#### Mid scale – a commitment





## Investment to secure some combination of:

- multiple diverse outcomes
- Speed
- Effective deployment of resources
- in some part of the business

# Mid-scale: the 4 pillars are expressed, and reproducible in some part of the business

People & know how

Group of Business leaders and IT Group of diverse Data Scientists Venues to manage the work

**Tech Stack** 

IT develops elements of modernized architecture, tech and possibly digital Data scientists have nascent data science platform

**Data** 

Data is on its way to FAIRification Reverse the 80-20 data effort allocation in some cases

Analytics & skepticism skills

Nascent formal process for analytical methodology: intake, prototype, validation, documentation, archiving and reuse

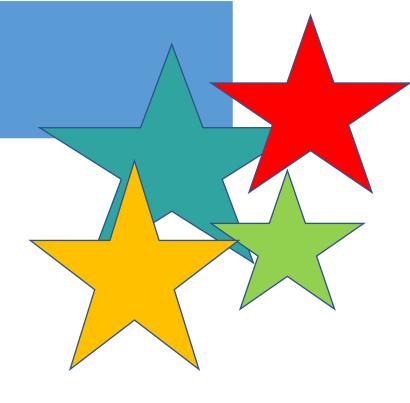
Modular investment proportional to the business needs and perceived value



## Mid-scale, learning is enhanced

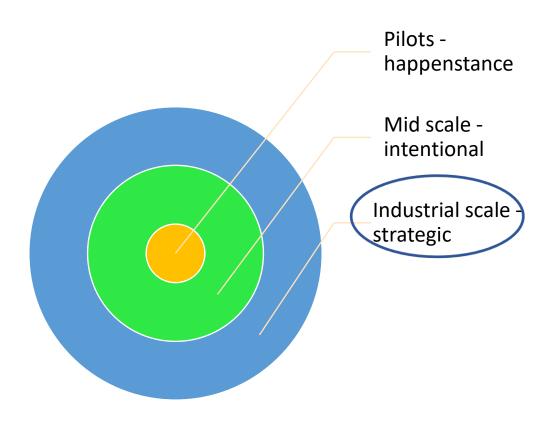
#### We learn

- How to organize larger groups of data scientists
- How to educate & translate, across IT, data science and business
- Build-test cycle of tech stacks
- Beyond hype, into the hard work of managing data. Roles, governance, cleanup...
- Analytics are actively managed



### Industrial scale – It's a way of life





#### BIG Investment to secure

- multiple diverse outcomes across the business
- Speed
- Evolution and enhancement of resources
- Future proof
- Culture

#### Industrial-scale



People & know how

Business leaders, IT, Data Scientists are aligned Group of diverse Data Scientists – managed across the enterprise Venues to drive insights uptake and monitor value

**Tech Stack** 

Modern and modular IT architecture, tech and digital Data scientists have data science platform to serve internal and external needs

Data

Data is managed as an asset, fully FAIR-ified. Can be prepared quickly and easily for multiple analytic approaches.

Analytics & skepticism skills

Established and monitored process for analytical work: apply analytics to analytics

### We are in the middle of a data (r)evolution

Many organisations are in the pilot or mid-scale stage

It takes A LOT to get to mid-scale, especially in older organisations

We have yet to see an industrial-scale implementation in healthcare

