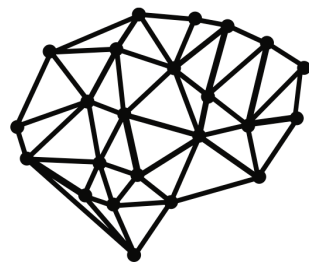


~~Вся правда~~ О специалистах по ~~данным~~ data science

Open
Data
Science



DM Labs

Алексей Натекин





Alex Natekin

SIEMENS

fortiss

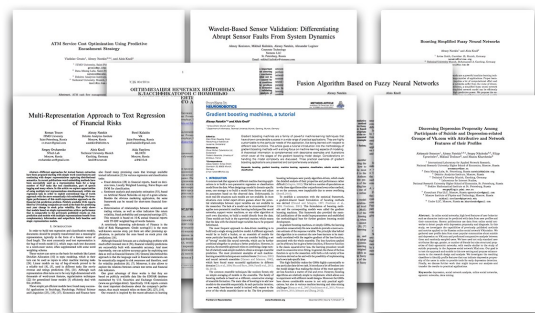
DM Labs

Deloitte.

DIGINETICA

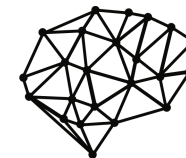


XOR



TUM TECHNISCHE
UNIVERSITÄT
MÜNCHEN

Open
Data
Science



2008

2009

2010

2011

2012

2013

2014

2015

2016

2017

Finding ~~nemo~~ data scientists:



Who are all
these people



How and where
to find them



Brief manual:
dos and donts

Data [driven] Science:

- ▶ “The art of turning data into actions”
- ▶ “Interdisciplinary field about scientific methods, processes and systems to extract knowledge or insights from data in various forms”
- ▶ “A newly emerging field dedicated to analyzing and manipulating data to derive insights and build data products”

Data Science

Here be data
visualization,
decision
theory,
domain
sciences,
optimization,
etc.

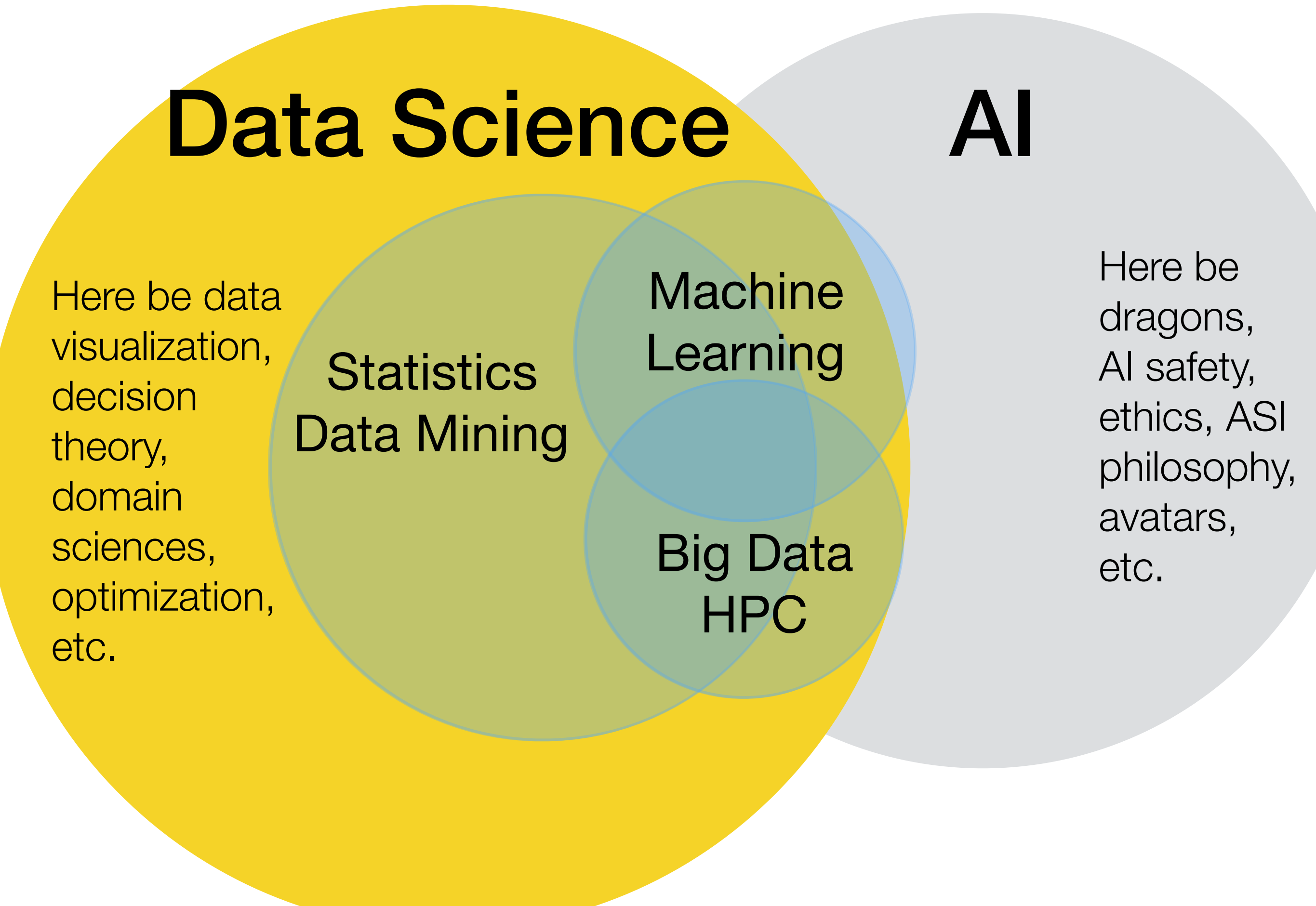
Statistics
Data Mining

Machine
Learning

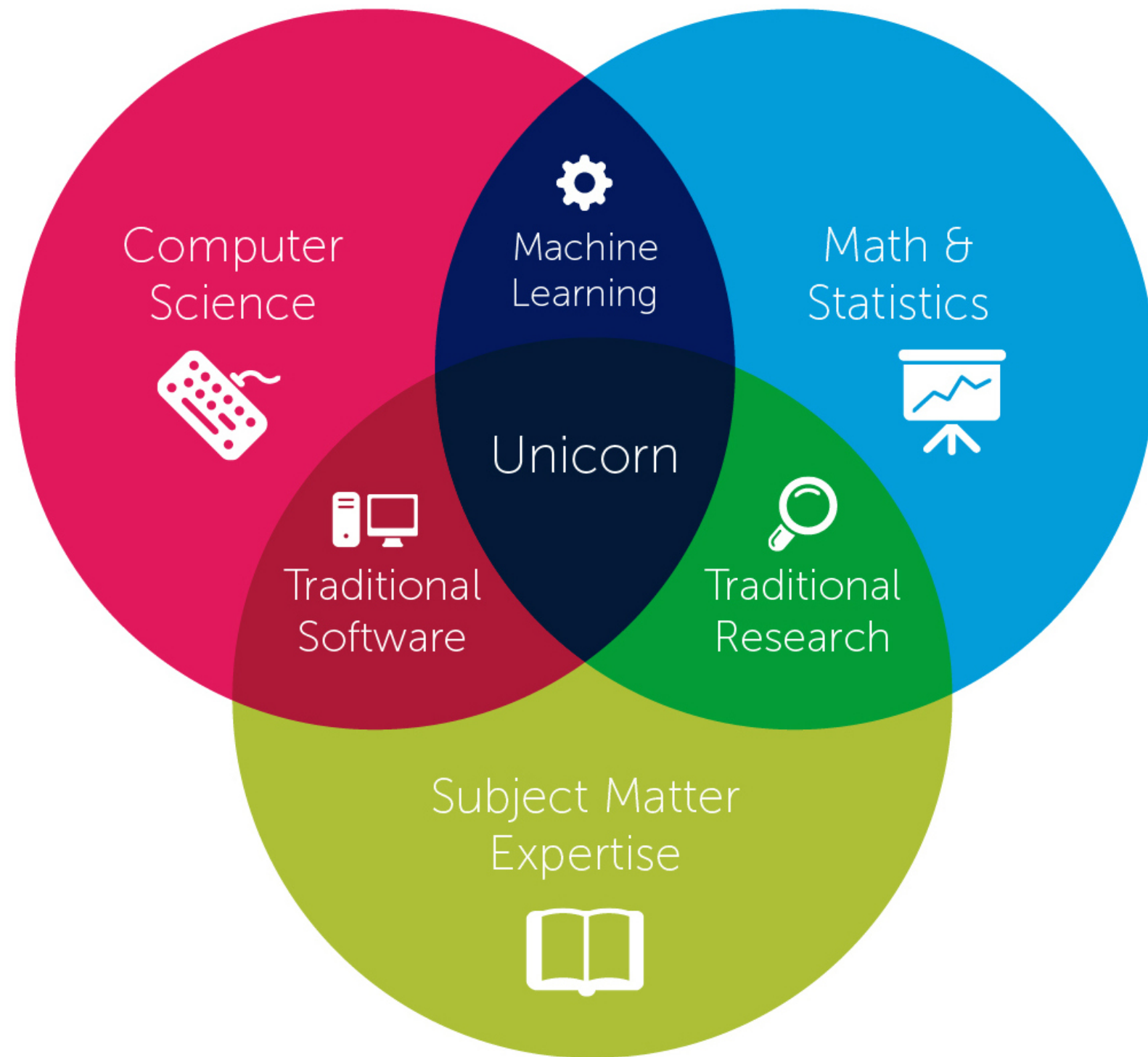
Big Data
HPC

AI

Here be
dragons,
AI safety,
ethics, ASI
philosophy,
avatars,
etc.



“A data scientist is someone who is better at statistics than any software engineer and better at software engineering than any statistician.”



Domain & business
expertise

Business & data
analysts

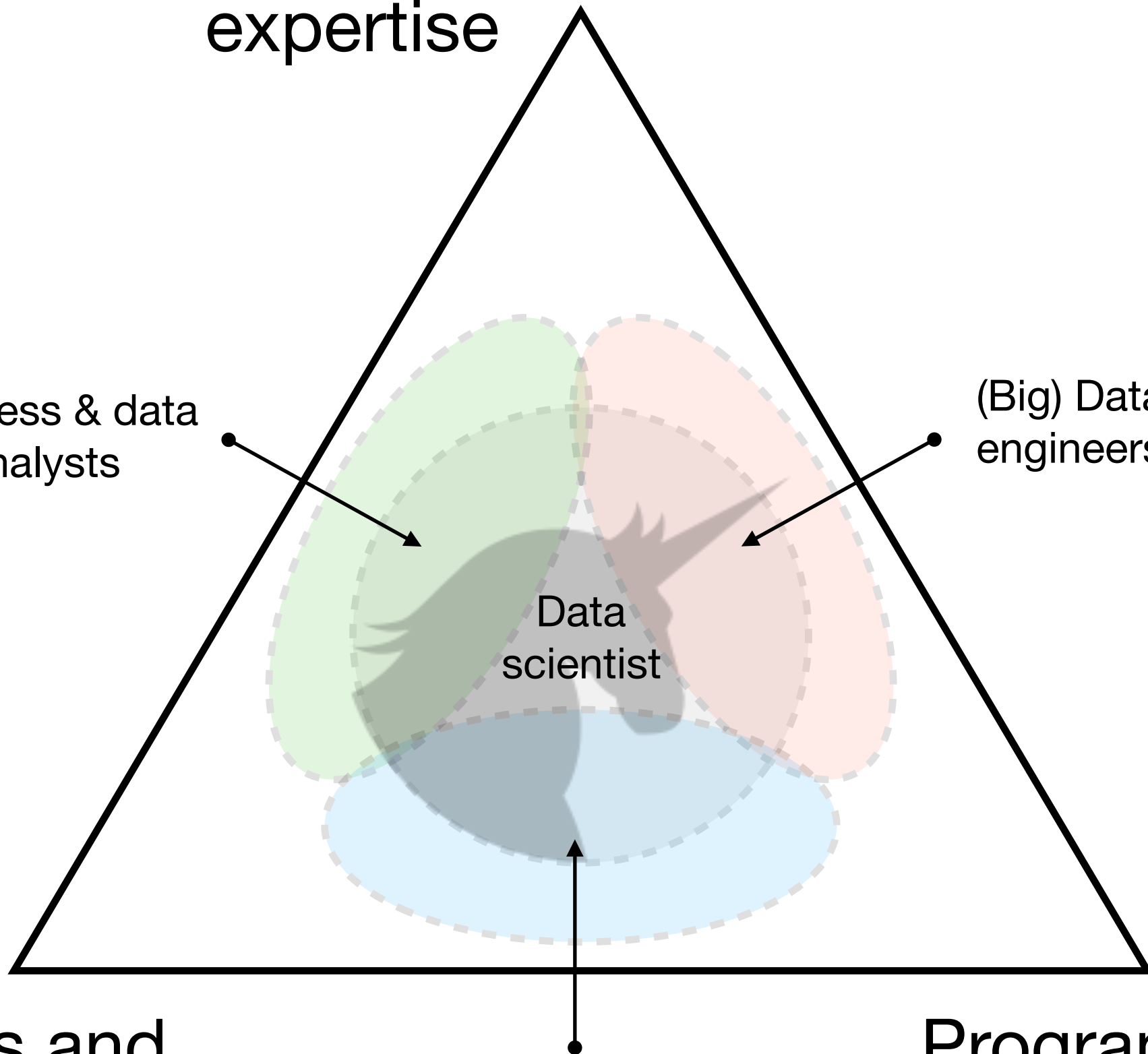
(Big) Data
engineers

Data
scientist

Statistics and
machine learning

(ML) research
scientists

Programming &
data engineering



Domain & business
expertise

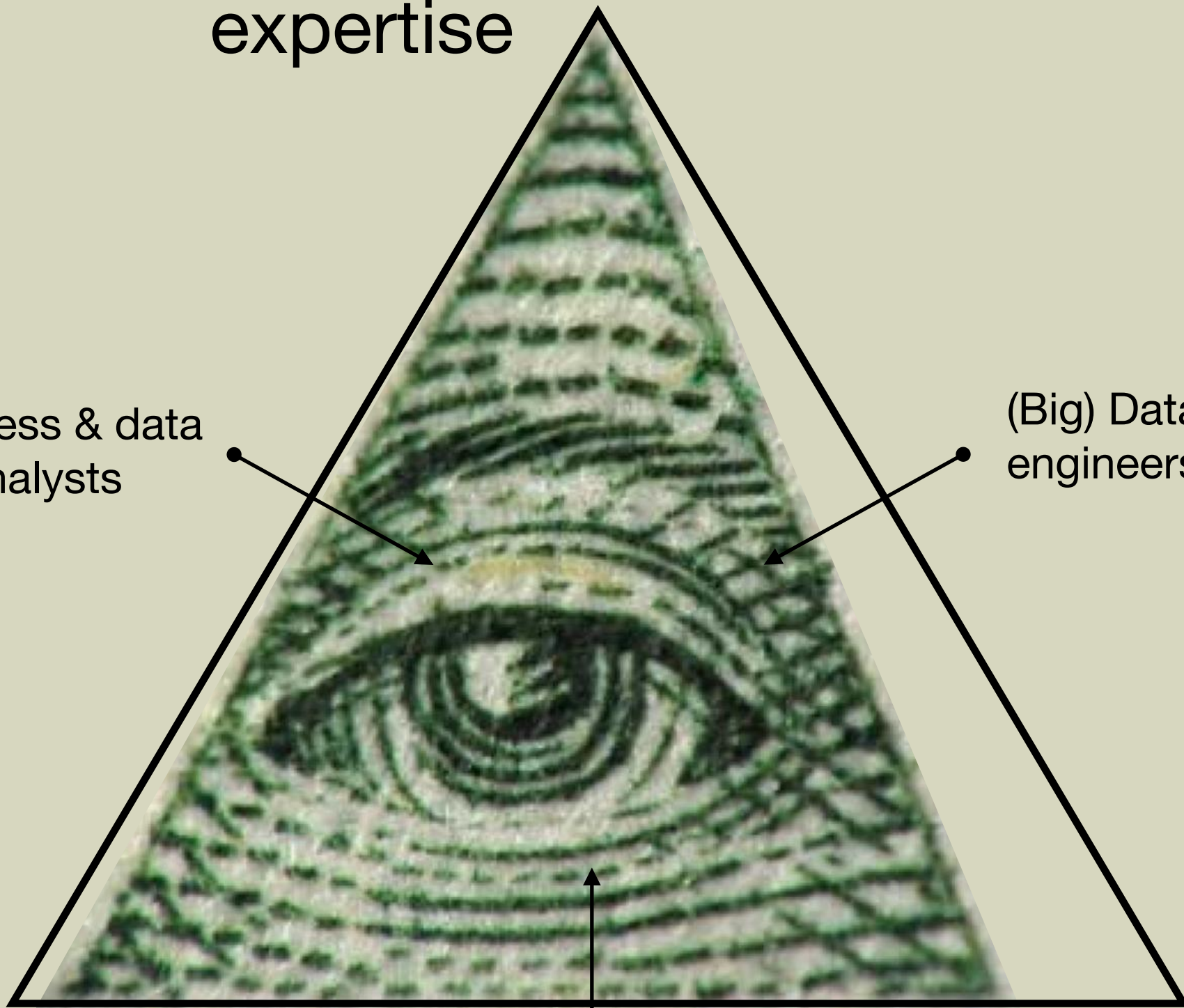
Business & data
analysts

(Big) Data
engineers

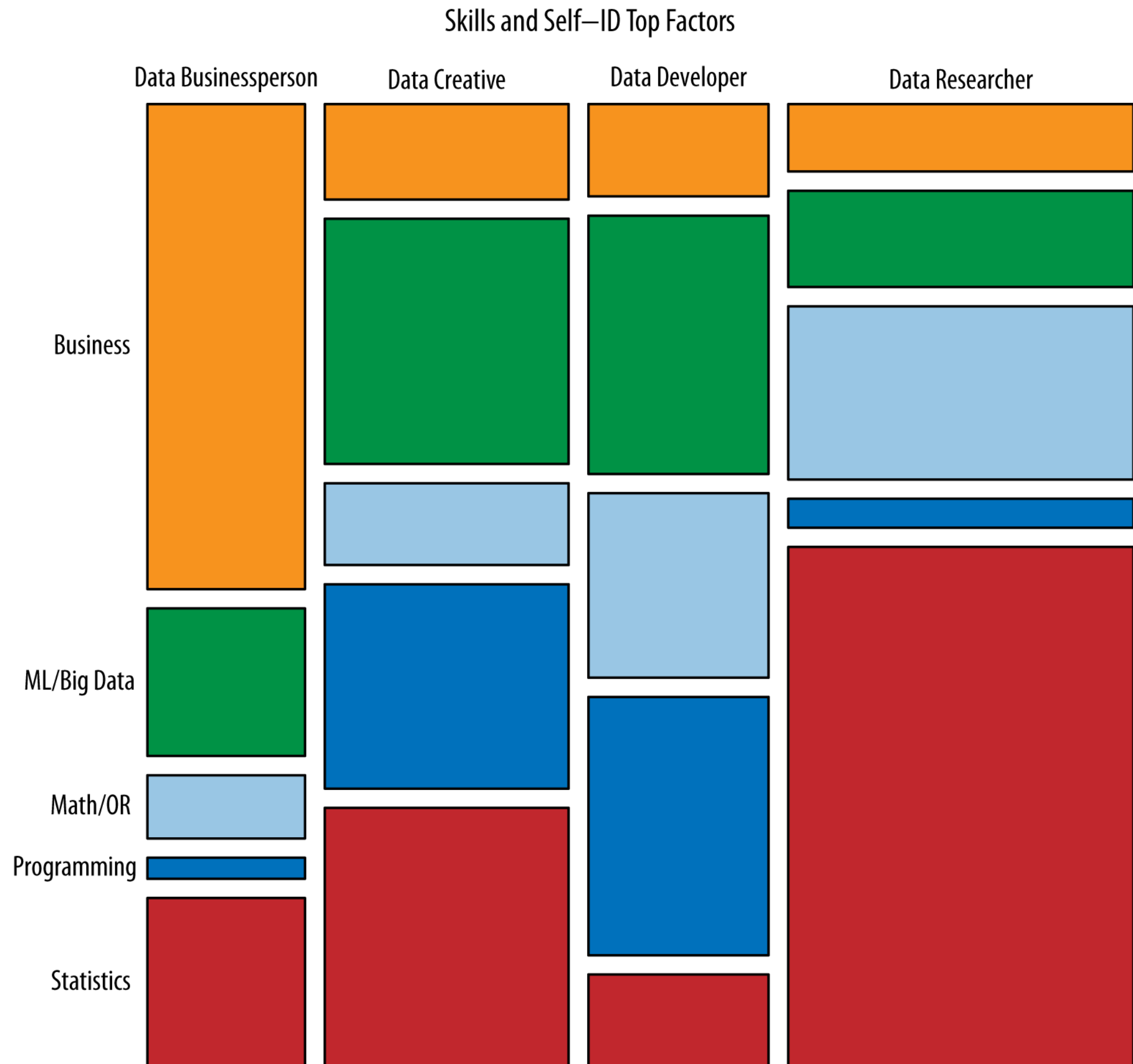
Statistics and
machine learning

(ML) research
scientists

Programming &
data engineering



Analyzing the Analyzers



Data scientists maturity:

Senior

You give them a (business) problem, they (consistently) bring you a (working) solution

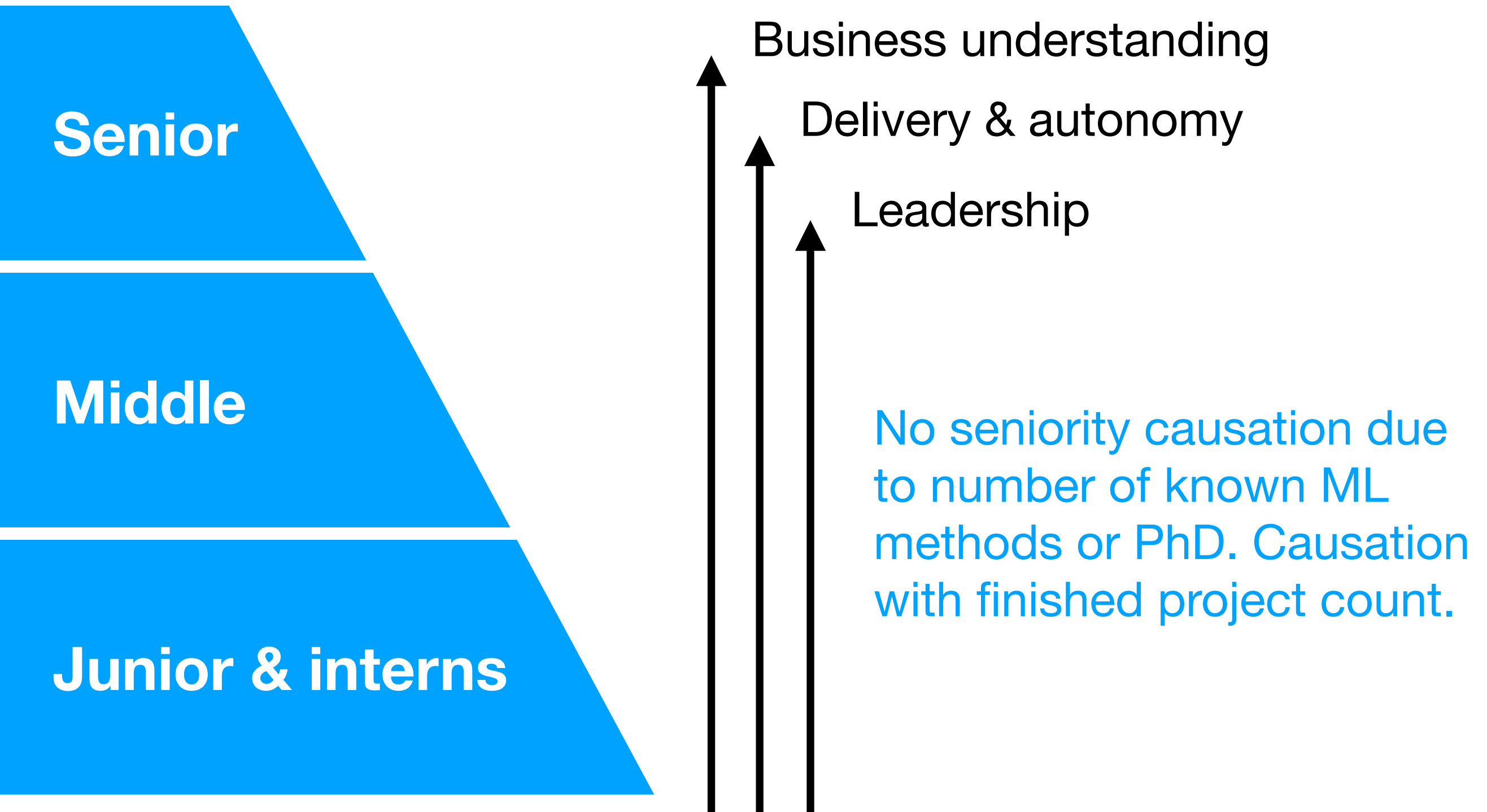
Middle

Given a prepared problem and paired with senior, you (weekly+) check on their progress

Junior & interns

Paired with mid and senior, they solve technical tasks and are checked daily

Data scientists progression:



Interview tips:

can be hacked

hard to hack

Senior

~~Statistics, ML, tech stack,~~
processes, development
skills **AND/OR** management,
business understanding

Middle

Statistics, ML, tech
stack, processes,
development skills

Junior & interns

Statistics, ML,
development
skills

Talk about
finished projects:

- business
- pet projects
- open source
- hackathons
- competitions
- study

All the projects, all
the details,
progression, why
use A and not B...



(1)

Senior

Large demand,
very low supply

Found via recruiters
Rare and few, easier
to validate

Middle

Some demand,
reasonable
supply

Found via job boards
& ~~recruiters~~. **Can be a**
too early junior promo

Junior & interns

Many
wannabe
in supply

Found via job boards
and academy. **Many**
impostors, hard to pick



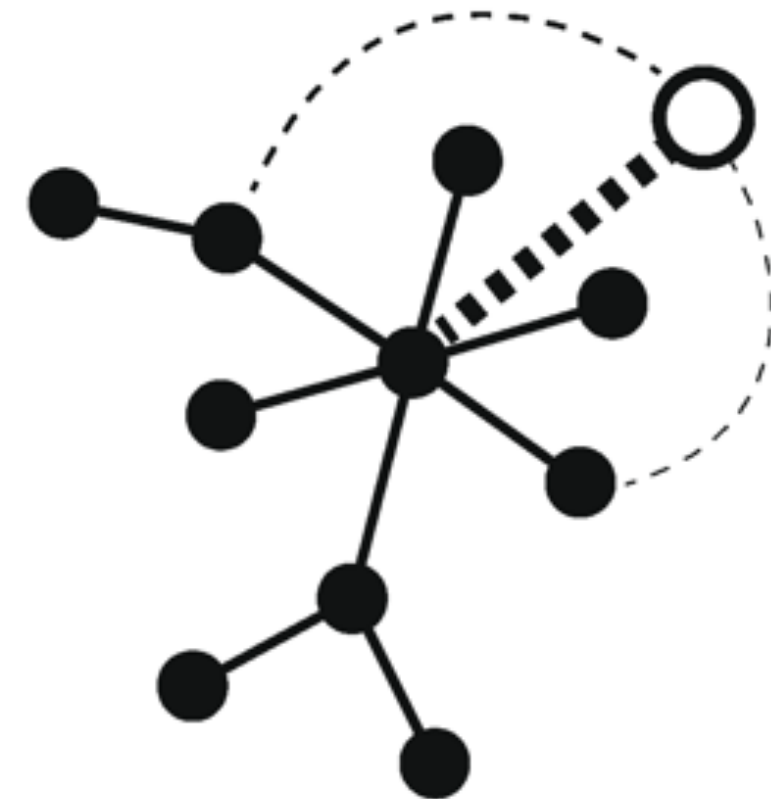
(2)

Senior

Middle

Junior & interns

Community
and personal
ties work best



Meetup

kaggle™

Data scientists specifics:



Notes about DS feasibility:

- Data science is not a silver bullet
Also, not all cases are reproducible and feasible
Don't overhype expectations
- Fail-safe ideology
Avoid a costly failure from the larger team
“If you fail, fail convincingly”
- Follow the “just do it” approach
Prototype and Succeed or Fail Fast
Help achieving more “80% solutions”

Notes about vacancies:

- Be precise with whom you are looking for
- If you want DS not to lie on an interview, don't lie on the job description. Give true salary estimates*
- Painful truth about actual tasks (ETL, no ML) is better than unmatched promises

*use java developers with same amount of experience and “lead” functions for proxy

Software and open source:





Common misconceptions:

- ▶ **Lack of vision in DS tasks**

- ▶ Expect DS team to operate as engineering team
- ▶ Confuse DS with pure data engineering role
- ▶ Fall for the seduction of tons of data

- ▶ **Forcing typical business patterns**

- ▶ Expect them to already know all the technologies, algorithms and approaches
- ▶ Overgrow teams: the bigger the slower
- ▶ Tell them what to do (like force them into agile*)

“We expect deep analytical talent shortage of between 140,000 and 190,000 by 2018 in the US only”

&

“We project a need for 1.5 million additional managers and analysts in the US who can ask the right questions and consume the results of the analysis of big data effectively.”

Summary:

- To prevent role mismatch, role of a data scientist requires more specification: data engineering, ml researcher, data analyst, domain expert, etc
- It's hard to find seniors, but juniors are also quite tricky. Use community and references for both
- Offer you data scientists a fail-safe and experiment-friendly environment
- Help DS with the right questions from business

Thanks!

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