

If you have questions during the presentation, you can ask me via slido.



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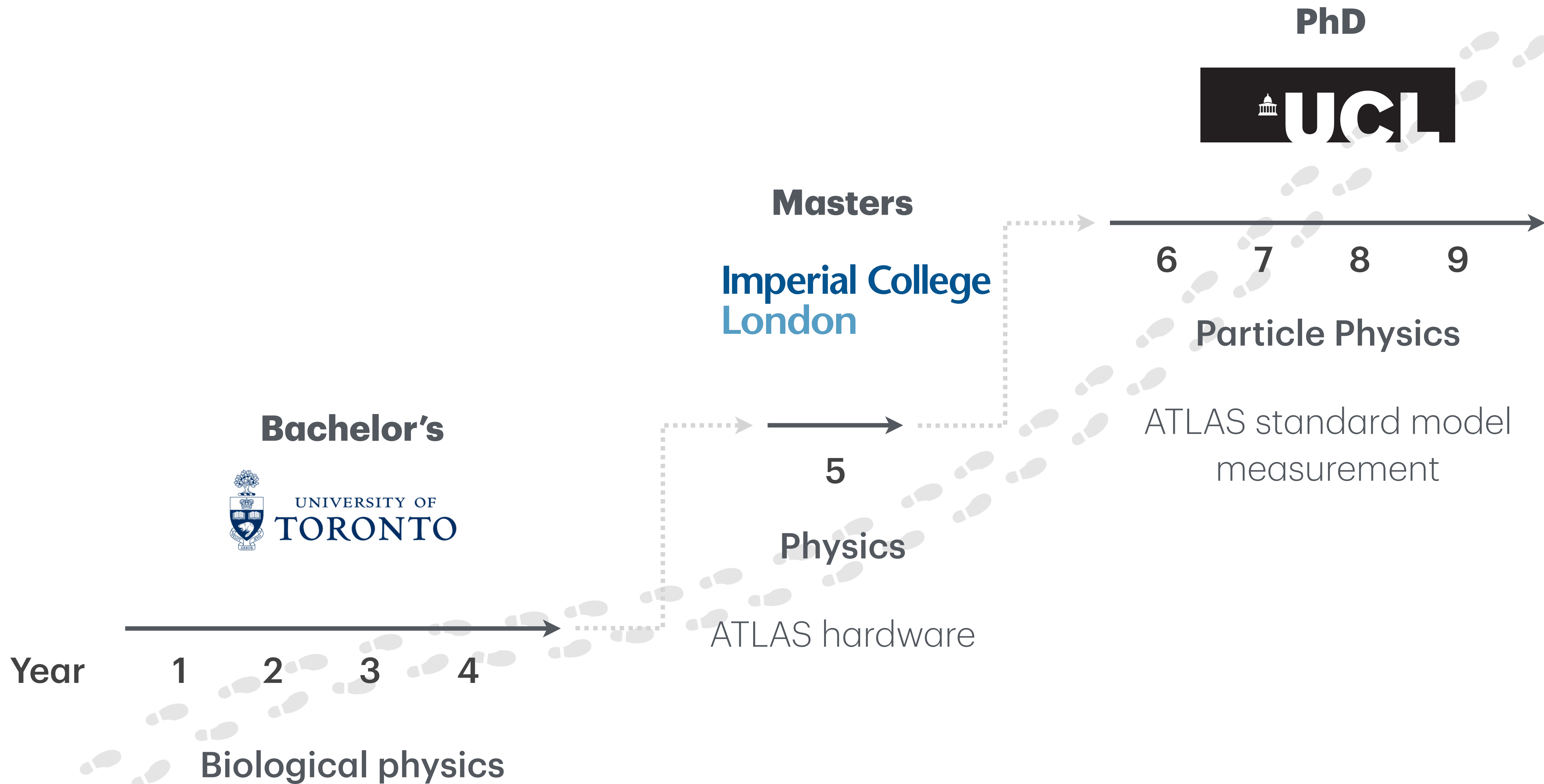
Journey from academia to industry

Joanna Huang

MCNet Summer School 2024

- 1 My academic background
- 2 Exploring industry career options
- 3 Identifying transferable skills
- 4 Bridging the gap
- 5 Navigating job applications and interviews
- 6 What is it like being a data scientist?
- 7 Lessons learnt and advice
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My academic background



Exploring careers in industry

What kind of jobs can I do?

Am I even qualified?

What if I don't like it?

Job hunting is so daunting, where do I start?

Will I be able to do the job?

What other skills do I need?



What kind of jobs can I do?



Research scientist

Machine learning engineer

Data analyst

Data engineer

Data scientist

What kind of jobs can I do?

*Are these all the
same jobs?!*

Research scientist

Machine learning engineer

Data analyst

Data engineer

Data scientist



What kind of jobs can I do?



Research scientist: advances knowledge through theoretical and experimental research.

Machine learning engineer: Designs, builds, and deploys machine learning models.

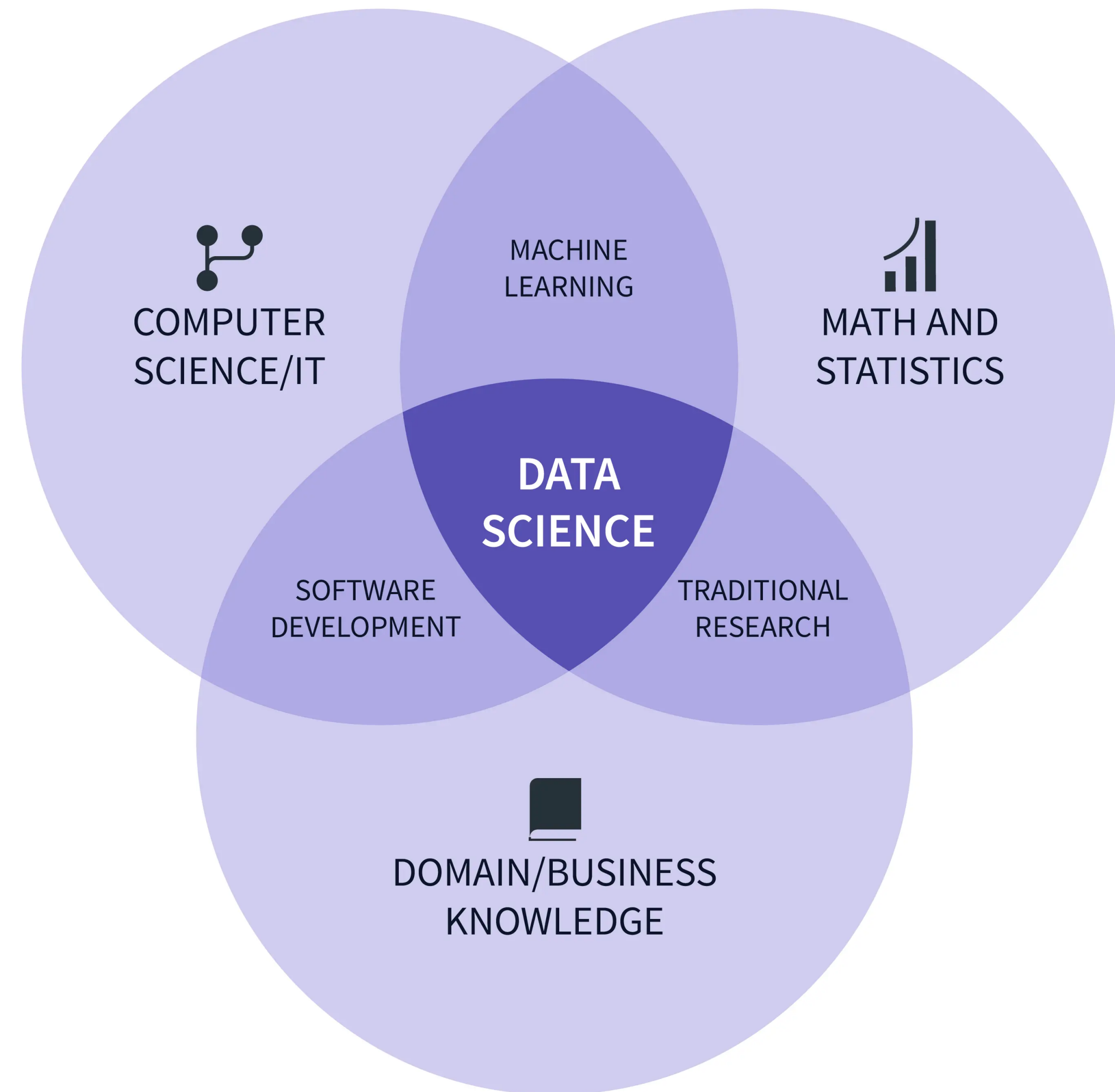
Data analyst: Analyses data to extract insights and inform decision-making.

Data engineer: Develops and maintains data architectures.

Data scientist: Applies statistical methods and machine learning to analyse data.

What kind of jobs can I do?

*Data scientist...
could that be me?*



***Data science** is an interdisciplinary field that lies at the intersection of several key domains*

Transferable skills

Technical skills

- Python (by far the most common language among data scientists)
- Using the command line (bash)
- Using version control, collaborative coding on gitlab
- Some back ground in statistics and maths



Soft skills


- Ability to work and problem solve independently
- Communication skills (ability to present and write explanations on complex topics)
- Having a scientific approach to problem solving




Bridging the gap

Data Scientist

London, England, United Kingdom · Reposted 2 weeks ago · Over 100 applicants

 Full-time · Entry level

 10,001+ employees · Insurance

Join us as an entry level data scientist! Where you will be doing some cool data science.

Required:

Python

Machine learning knowledge

Previous data science experience

Nice to have:

PhD in machine learning, statistics, maths, or physics

MLOps

SQL

Deep learning/NLP/computer vision


Cloud technologies



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*That's quite a few
knowledge gaps*





I have little to no machine learning expertise... what can I do to change that?

General advice

- **Hunt for Opportunities:** Look for fellowships, secondments, and other learning opportunities.
- **Online Courses:** Take advantage of online courses to build foundational knowledge.

➔ Coursera, edX, Udacity, and Khan Academy

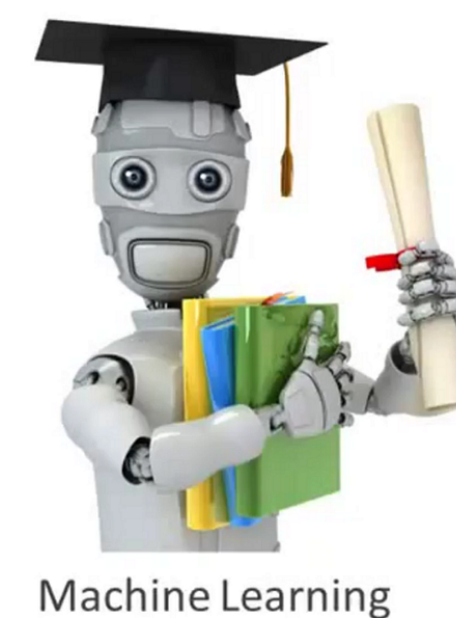
<p>Machine Learning Specialization</p> <p>Multiple educators</p> <p>Machine Learning</p> <p>Skills you'll gain: Machine Learning, Machine Learning Algorithms, Applied Machine Learning, Algorithms, Deep Learning,...</p> <p>Make progress toward a degree</p> <p>★ 4.9 (23K reviews)</p> <p>Beginner · Specialization · 1 - 3 Months</p>	<p>University of Washington</p> <p>Machine Learning</p> <p>Skills you'll gain: Machine Learning, Algorithms, Machine Learning Algorithms, Human Learning, Applied Machine Learnin...</p> <p>★ 4.6 (16K reviews)</p> <p>Intermediate · Specialization · 3 - 6 Months</p>	<p>IBM</p> <p>Machine Learning with Python</p> <p>Skills you'll gain: Machine Learning, Algorithms, Data Analysis, Human Learning, Python Programming, Regression</p> <p>★ 4.7 (16K reviews)</p> <p>Intermediate · Course · 1 - 3 Months</p>	<p>Deep Learning Specialization</p> <p>DeepLearning.AI</p> <p>Deep Learning</p> <p>Skills you'll gain: Machine Learning, Deep Learning, Artificial Neural Networks, Machine Learning Algorithms, Applied Machine...</p> <p>Make progress toward a degree</p> <p>★ 4.8 (143K reviews)</p> <p>Intermediate · Specialization · 3 - 6 Months</p>	<p>Google Cloud</p> <p>Introduction to Generative AI</p> <p>Skills you'll gain: Artificial Neural Networks, Deep Learning, Machine Learning</p> <p>★ 4.7 (4.5K reviews)</p> <p>Beginner · Course · 1 - 4 Weeks</p>
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I have little to no machine learning expertise... what can I do to change that?

My path

- 3 month data science fellowship (via UCL project based)
- 1 month data science fellowship (online, courses based)
- Online machine learning courses
- Joined machine learning communities on Kaggle and LinkedIn



Machine Learning

by Andrew Ng



Navigating applications and interviews

The most daunting part about the transition is the application and interview process...



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Tips for applications

- **Tailor your resume**
 - Get feedback from peers
 - Incorporate relevant keywords from the job description
 - Highlight and quantify your achievements



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Tips for applications

- **Tailor your resume**
 - Get feedback from peers
 - Incorporate relevant keywords from the job description
 - Highlight and quantify your achievements
- **Network, Network, Network!**
 - Leverage LinkedIn
 - Attend events
 - Seek out referrals



The most daunting part about the transition is the application and interview process...

What to expect for interviews

- **Screening call**
 - Are you a good fit for the role? Is the role what you are looking for?
- **Coding test**
 - Automatically scored coding tests
 - Live coding tests
- **Take home assignment**
 - Comprehensive project that you submit after a few days

The most daunting part about the transition is the application and interview process...

Tips for interviews

- **Research the company**
 - What are their core values? What is the product or service provided?
- **Practice coding under pressure**
 - Use platforms like LeetCode to practice solving problems within a time limit
 - Simulate live coding tests by solving problems on a call with a peer.
- **Practise building models using data from Kaggle**
 - Create a Jupyter Notebook as the final submission report.
 - Provide clear documentation of your approach and note any limitations or constraints

After nine years of studies, I took the leap and transitioned from physics to data science.

faculty



What is it like being a data scientist?

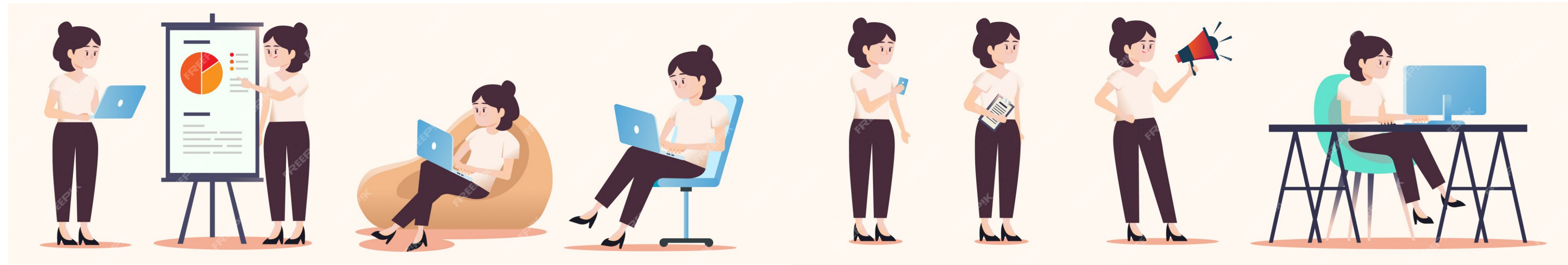
As a data scientist, my main responsibility is focused on solving real-world problems through machine learning

My key responsibilities on a consulting project as a data scientist are:

1. Exploratory data analysis.
2. Translating client needs into a technical problem that can be solved with machine learning, taking data restrictions into account.
3. Designing and implementing bespoke ML solutions.
4. Prepare the solution as a Python package.
5. Support engineers with the deployment[💡] of the solution



A day in my life as data scientist looks something like this



- **Daily “stand-up” meeting** first thing in the morning with my team
 - Did I do what I said I would do yesterday? If no why not? What is my plan for today? Do I have any blockers?
- **Focus time** to do some work
 - Usually involves writing code, using Jupyter notebooks, lots of merge requests, documenting results and sharing that with the team
 - Lots of “pairing sessions” where you pair code with a technical colleague to solve problems together
- **Meetings**, meetings, and more meetings
 - Commercial meetings, data science meetings, engineering meetings, company-wide meetings, project-specific meetings, knowledge share meetings, development meetings

Key differences between academia and industry

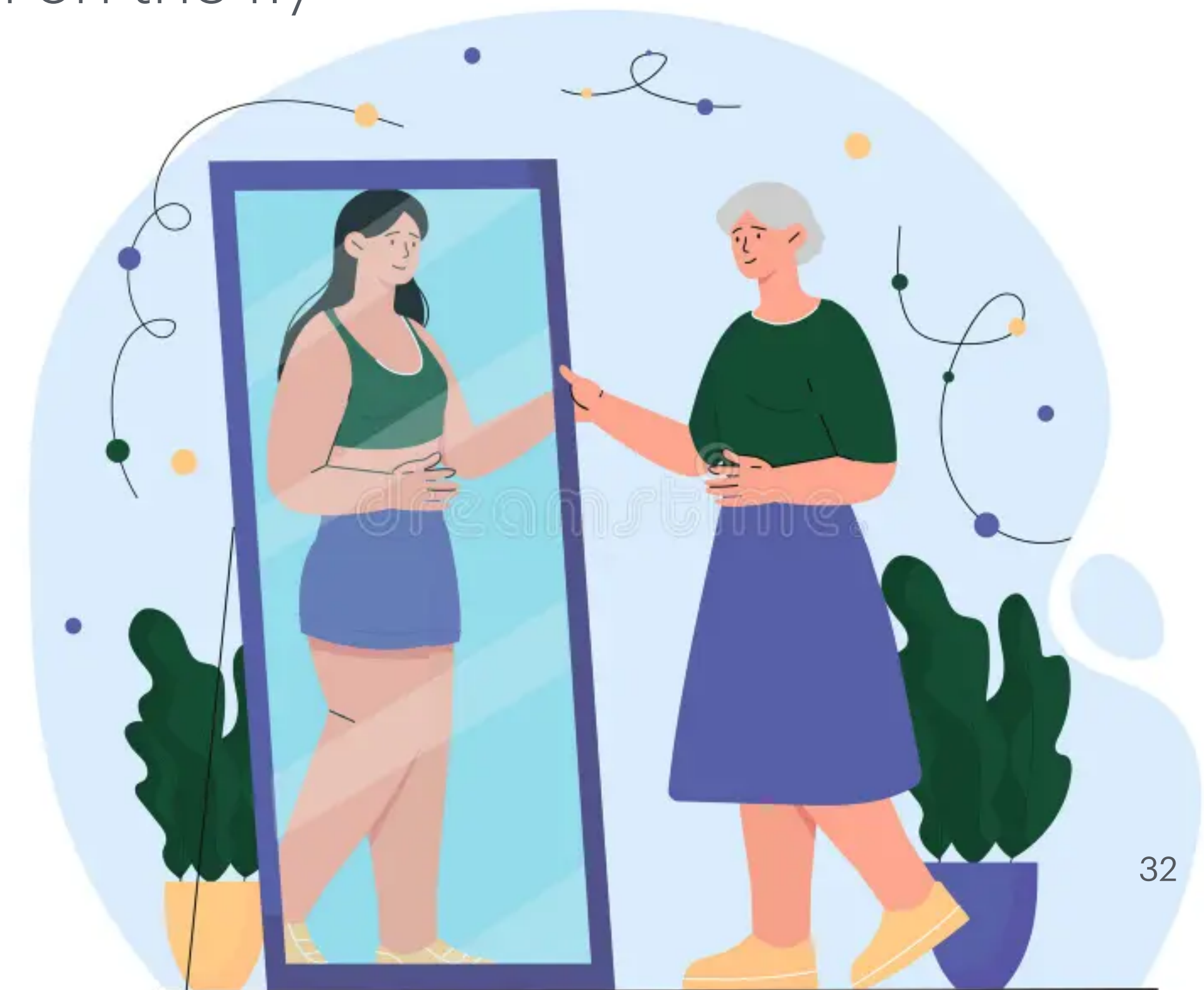
- Academia is driven by personal **research goals**, while industry is driven by **corporate goals**
- In academia, research is driven by the **pursuit of knowledge**, whereas industry focuses on applying research for **practical impact**
- Academia offers **flexible schedules**, while industry follows set **business hours**
- Academia involves **independent working**, while industry is more **collaborative**
- Academia offers competitive and **limited career paths**, while industry provides **diverse opportunities and stability**



Lessons learnt and advice

Insights and wisdom I would share with my past self

- Your PhD prepares you more thoroughly for industry roles than you think
- Checking every single box in the job description isn't necessary
- Whatever gaps there are in your knowledge, you can learn on the fly
- Don't underestimate the importance of networking
- Approach interviews with the same diligence as studying for exams, your job depends on it
- Recognise and confidently showcase your true value; don't underestimate yourself!



My advice to academics looking to transition into industry

- **Develop transferable skills:** Focus on communication and teamwork alongside technical expertise.
- **Network actively:** Take advantage of your professional networks. Reach out to your peers for advice
- **Research industry needs:** Understand industry demands to tailor your resume to them.
- **Gain practical experience:** Seek internships, fellowships, freelance work, etc. to boost your resume.
- **Stay flexible and adaptable:** Embrace uncertainty and be open to exploring different career paths.
- **Invest in continuous learning:** Technology evolves quickly so seek out ways to keep yourself up to date.



Q&A



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