

# Advancing Careers: Grad School & Jobs

## *Week 3*

**DS198-003: Data Discovery Scholars Seminar**  
*UC Berkeley - Computation, Data Science, and Society*

*Spring 2022*

# Meme of the Week

ARE YA hired  
SON??



NO.



# Announcements

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- Please complete the following
  - Mid-semester survey: [tinyurl.com/ds198-mid](https://tinyurl.com/ds198-mid)
    - Thanks to those who've already completed it
  - Career Accelerator Program form: <https://forms.gle/C1pG8YsvfGpvvgPhi9>
    - Complete if you are interested. Please do so by 6:30PM today.
- Mid-semester presentations next week

# Tech Recruitment

Software Engineering

Product Manager

Data  
Scientist

**What job do I apply to?** 🤯

Data  
Analyst

Business Analyst

ML Engineering

Data Engineering

# Tech Jobs

## Business Analyst

- Improves business processes between the business and IT
- MS Office, SQL, Data Visualization, business intelligence, understanding, modeling

## Software Engineer

- Building Products, Real Time API, Servers, Cloud Computing, Automation, Mobile/ Web Applications
- Java, Ruby, Python, Swift, JS, Go, C, C++, C#,

## Product Manager

- The CEO of a product
- Asana, JIRA, Tableau, General Data Analysis,
- Product Analysis, project management, interpersonal skills, elementary technical skills to step in



# Tech Jobs

## ML Engineer

- Developing universal algorithms to extract meaningful information from large amounts of data
- C++, Python, Java or R
- Deep Learning, Heavy Probability, Statistics, Linear Algebra

## Data Analyst

- Collects, process, and performs statistical data analyses
- Excel, SQL, Tableau, Power BI
- Math, Stats, Machine Learning

## Data Engineer

- Develops, constructs, tests and maintains architecture
- SQL, Hive, Pig, R, Hadoop, Spark, Matlab, Python, Java, Ruby, C++
- Database systems, modeling, APIs

## Data Scientist

- Cleans and organizes big data from the POV of expertise
- R, SAS, Python, SQL, Data Science Toolkit
- Predictive modeling, story-telling
- Math, Stats, Machine Learning

**Check the job description!**

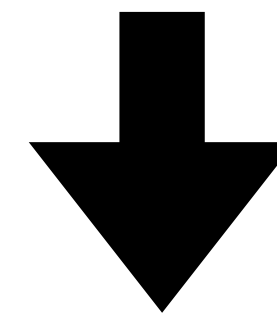
# Application Styles

## Spray & Pray

- Same resume
- Boring to no cover letter
- No networking
- No company research
- **High** number of applications

## Tailored Approach

- Targeted resume
- Individual cover letter
- Crazy networking
- **Low** number of applications



## Hybrid

- Resume for different jobs
- Cover letter template
- Networking when possible
- **High** number of applicants

# Networking

- **Linkedin**
- **Phone Chats/Zoom Calls**
  - Acquaintances, alumni, co-workers, bosses
- Question to ask
  - Do you have advice for someone who is in this field?
  - What do you do at your company?
- You are not bothering that person! Networking is beneficial for both.



# Resources: Applying to Jobs

- Tech Jobs
  - <https://skillcrush.com/blog/41-tech-job-titles/>
  - <https://towardsdatascience.com/data-scientist-vs-machine-learning-engineer-skills-heres-the-difference-93eb2f4f6f98>
- Resume
  - <https://career.berkeley.edu/Tools/Resume>
  - <https://www.themuse.com/advice/how-to-make-sure-your-resume-is-as-current-as-your-skills>
- Coffee Chat/Networking
  - <https://www.indeed.com/career-advice/finding-a-job/how-to-network-for-a-job>

# The Interview

- **Phase 1: Coding/Take Home Challenge**

- Goal: Assessing general programming knowledge as a way to filter out unfit candidates
- *Time:* 30 minutes to 3 hours

- **Phase 2: Technical Phone Screen**

- Goal: Screening a combination of technical, interpersonal skills and history
- *Time:* 30 minutes to 1 hour

- **Phase 3: Onsite Interview & Whiteboarding**

- Goal: Looking for coding skills, problem-solving ability, creativity, culture fit, analytical thinking, handling feedback, understanding the bigger picture
- *Time:* 1 hour to a day

# The Interview: Differences

- Though the previous slide is a general overview, there are wide variations in interview setups
- **Startups:** Tend to have shorter interviews that exist of behavioral, SQL coding tasks and resume questions.
- **Data Science:** Varies widely. Take-home exercises, SWE interview, statistics, behavioral/resume questions.
- **Machine Learning:** Combination of Data Science & SWE interview with heavy emphasis on job-related skills (think: CV, NLP, RL)
- **Data Engineering:** Combination of SWE & Systems Design questions.
- **Business Analyst:** Light SWE interview and focus on business topics and drawing conclusions.





# The Interview: Prepare

- CS61B & DATA100 are your best friend
  - Data Structures, Big O, Trees, Sorting, Heaps, Hashing
  - Domain Specific (e.g. DL for ML Engineering)
  - System Design
  - Condensed forms can be found online (Cracking the Coding)
- Practice your riddles: LeetCode, HackerRank
  - Remember, there is only a finite number of problems they can give you
- Review your projects
  - Do you remember to explain all your decisions?



# Tips

- Think out loud while solving problems
  - The interviewer cannot pick your brain
  - Explicitly mention each step and the motivation behind it
- **Know your audience**
  -  Bragging about your AUC score in front of a recruiter
  -  Covering up a big flaw in your project in front of an engineer
- Don't lie, don't pretend, focus on what you do know
- Do your research

# The Interview: Resources

- <https://github.com/yangshun/tech-interview-handbook>
- <https://github.com/datastacktv/data-engineer-roadmap>
- <https://github.com/khangich/machine-learning-interview>
- <https://www.crackingthecodinginterview.com>
- <https://www.kdnuggets.com/2021/08/common-data-science-interview-questions-answers.html>

# Graduate School

# Do I need a Graduate Degree?

- Ask yourself the following **questions**:
  - What do you want to get out of the program?
  - Is it worth the money and labor?
  - Do I have the right background and preparations?

For instance, you want to:

- become a professor
- get a higher job prospect
- are hunting for specific niche jobs
- become more employable
- get more depth



# Master's vs. PhD

## Master's

- 1-2 years
- Can do research and teaching, but mostly course based
- Not always funded
- Breadth
- Diverse population
- Normally enough if you don't want to go into academia

## Doctoral

- 4-5 years
- Teaching & Research-based
- Requires pre-lims and quals
- Funded
- Depth & Expertise
- Normally needed for teaching or certain specific jobs

# Applying to schools

- Factors to take into considerations
  - **Research Field**
  - **Course**
  - **Program Length**
  - **Cost**
  - **Prerequisites & Competitiveness**



# The Essay: SOP vs. PS

- The **Statement of Purpose** is mostly an academic document about:
  - What you did, you want to do during graduate school and afterwards **research/course-wise**
- The **Personal Statement** is more like the college essay. It asks about your personal story and how that ties in with your ambition/passion.
- **Strategy: Applying to multiple programs**
  - Start with the school with the most specific prompt. Then match and edit it accordingly for other schools.

# Other Tips

- **Letter of Recommendation**

- Ask a letter from a professor, teacher or manager that knows you well and has worked with you rather than a distant letter from a famous professor you have seen once.
  - A good letter from an unknown person is much better than a bad letter from a famous person

- **GRE**

- There are free resources to prepare for the GRE online (Magoosh, ETS)
- Many schools have gotten rid of the GRE

- **Extra Help**

- Many schools provide resources for underrepresented students to help with their applications
- Ask your friends to go over your essays!



# Resources

- **Choosing a program**

- <https://www.gograd.org/resources/choosing-graduate-program/>
- <https://www.cs.cmu.edu/~pavlo/blog/2015/10/how-to-write-a-bad-statement-for-a-computer-science-phd-admissions-application.html>

- **GRE**

- [www.reddit.com/r/GradSchool/comments/1kvk0f/best\\_way\\_to\\_prep\\_for\\_gre/?usg=AOvVaw2tctrj6YeF-alL6n72qsN\\_](http://www.reddit.com/r/GradSchool/comments/1kvk0f/best_way_to_prep_for_gre/?usg=AOvVaw2tctrj6YeF-alL6n72qsN_)

- **Essay**

- <https://www.cs.cmu.edu/~pavlo/blog/2015/10/how-to-write-a-bad-statement-for-a-computer-science-phd-admissions-application.html>
- <https://career.berkeley.edu/Grad/GradStatement>

# Resources (cont'ed)

- **Programs for URM**
  - <https://github.com/gwisk/gradguide>
  - <https://github.com/chinasaokolo/csGraduateApps>

# Data Science Interview

# Data Science Interview

**Form pairs. One person will be the interviewer, the other person the interviewee. The interviewee has 5 minutes to elaborate their thought process on the first question/case. After 5 minutes, switch roles and proceed to the second question/case.**

## Case/Questions

- (1) You work as a data scientist for a ride-sharing company. An executive asks how you would evaluate whether a 50% rider discount promotion is a good or bad idea. How would you implement it? What metrics would you track?
- (2) We're working on a new feature for LinkedIn chat, and we want to implement a green dot to show an “active user”. Given engineering constraints, we can't AB test it before release. How would you analyze the effectiveness of this new feature?

# Check-In

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- No class next week! Instead, we will be conducting 15-20 min one-on-one check-ins via Zoom. This meeting serves to:
  - get to know you better
  - discuss how you are feeling overall
  - talk about how the project is going
  - plan ahead your project
  - answer any questions you have related to your project

# Work Session & Office Hours