

A decorative graphic in the top-left corner consisting of two overlapping parallelograms: a blue one in the foreground and a light green one behind it, both slanted downwards from left to right.

Data Science Skills

Team Tidy

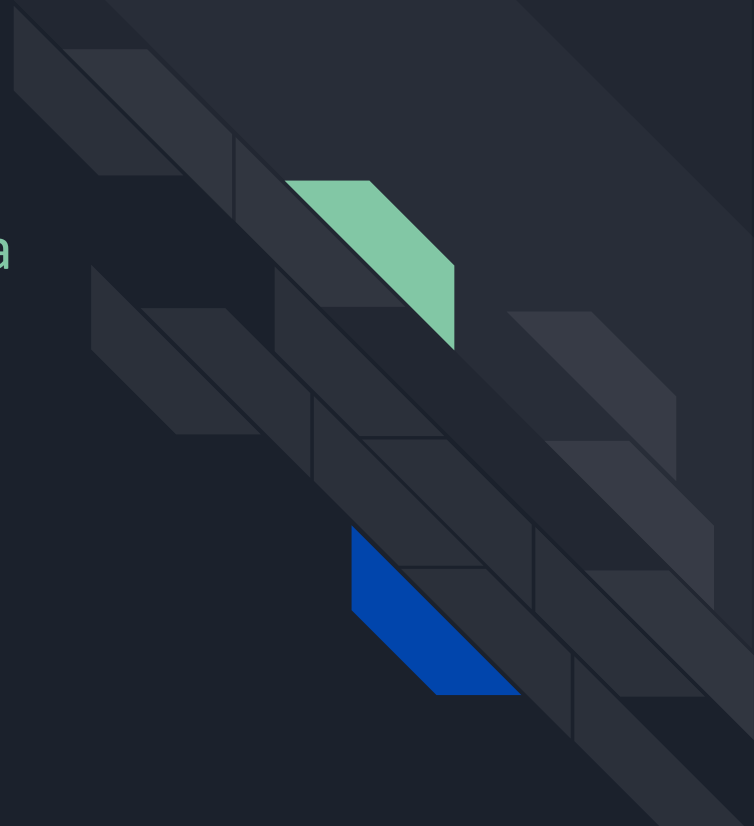
Alec McCabe

Chinedu Onyeka

Cliff Lee

Preston Peck

Santiago Torres





Agenda

Section	Team Member	Time
Approach <ul style="list-style-type: none">• Tools• Assumptions	Santiago	1 minute
Data Collection <ul style="list-style-type: none">• Web Scraping• Persistent Storage	Alec & Cliff	3 minutes
Data Transformation <ul style="list-style-type: none">• EMSI	Preston	1 minute
Data Analysis	Chinedu & Preston	2 minutes
Conclusion	Santiago	1 minute

Approach





Approach

For DATA 607 Project 3, all teams must use data to answer the question, “Which are the most valued data science skills?” Consider your work as an exploration; there is not necessarily a “right answer.”

Gather job postings and survey what the job market finds valuable



Collaboration Tools

Communication

- Slack
- Zoom

Project Documentation

- Google Drive

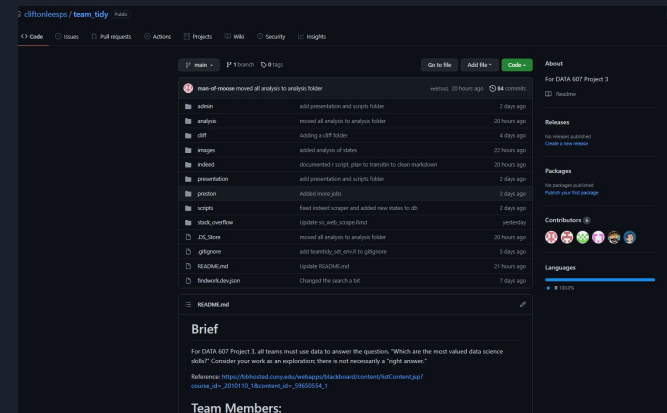
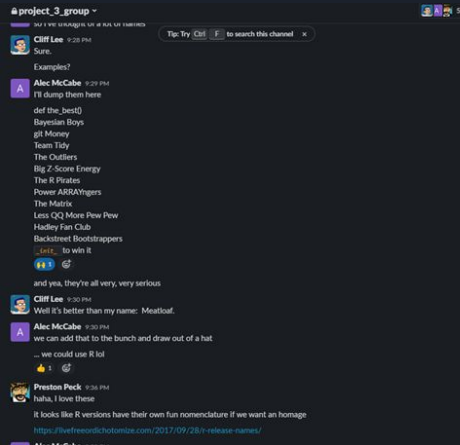
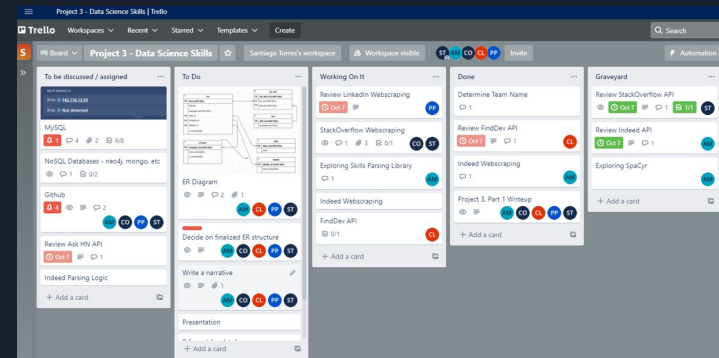
Project Management

- Trello

Code Sharing

- GitHub

https://github.com/cliftonleesps/team_tidy

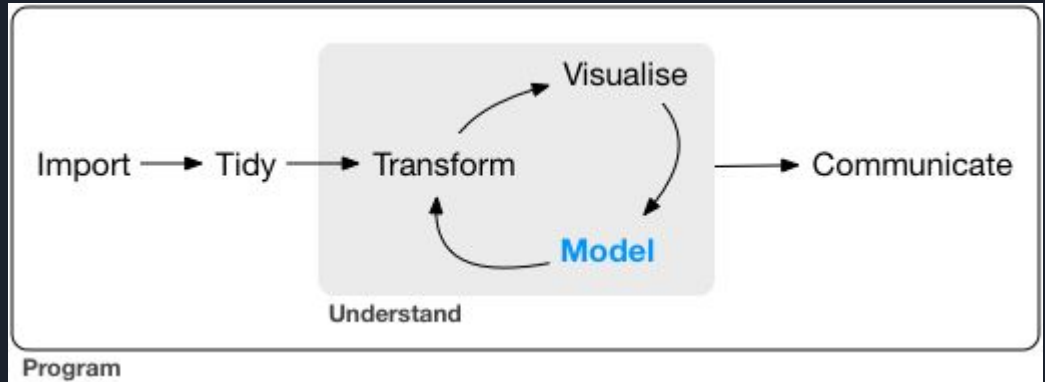


Assumptions

We'd see skills that fall along equally on each part of the model:

- Import / Tidy / Transform skills
- Visualization skills
- Modeling skills
- Communication skills

Hadley's Data Science Model



Data Collection

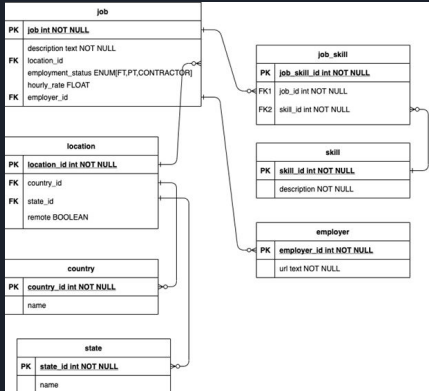


Persistent Storage



Job Title	State	Company	Original Source	Description	Type
IT Data Scientist	California	JPL/NASA	indeed	Basic Math	Soft Skill
				Communications	Soft Skill
				Creative Thinking	Soft Skill
				Creativity	Soft Skill
				Curiosity	Soft Skill
				Innovation	Soft Skill
				Management	Soft Skill
				Mentorship	Soft Skill
				Presentations	Soft Skill
				Problem Solving	Soft Skill
				Professionalism	Soft Skill
				Resourcefulness	Soft Skill
				Self Starter	Soft Skill
				Self-Awareness	Soft Skill
				Agile Methodology	Hard Skill
				Analytics	Hard Skill
				Application Specific Integrated Circuits	Hard Skill
				Auditing	Hard Skill
				Business Operations	Hard Skill
				Business Process	Hard Skill
				Computer Science	Hard Skill
				Data Analysis	Hard Skill
				Data Integration	Hard Skill
				Data Mining	Hard Skill
				Data Modeling	Hard Skill
				Data Quality	Hard Skill
				Data Science	Hard Skill
				Data Visualization	Hard Skill
				Data Wrangling	Hard Skill
				Machine Learning	Hard Skill

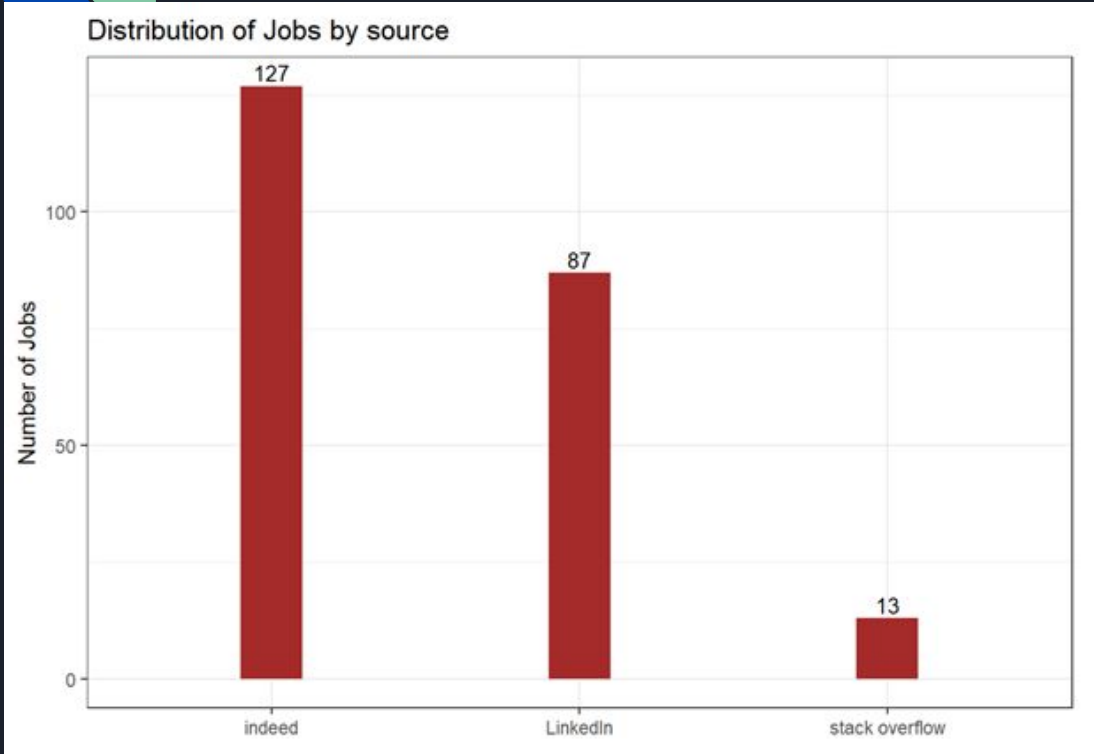
ER Diagram



job	skill
PK job_id int NOT NULL	PK skill_id int NOT NULL
job_title text	description text NOT NULL
description text NOT NULL	job_id int NOT NULL (FK)
state	type [hard, soft]
country	
company_name	
company_url	
url varchar(255)	
employment_type	
min_salary	
max_salary	
original_source	

Sample Data (Soft Skills in Red)

Web scraping



Web-scraped 220 jobs from:

- Indeed
- LinkedIn
- StackOverflow

Technologies used:

- Xml2
- Rselenium

Similar structure between websites:

- Jobcards collected from page
- Iterate through pages

Data Transformation



Data Transformation

Extracting meaning

- With all our **job description text**, how do we extract the key **skill data** effectively and efficiently?
- We could parse manually or hope for predictable patterns to parse by, but this is tedious and inconsistent
- We explored options for Natural Language Processing (NLP) services with Named Entity Recognition (NER) models...

prodigy

Text Classification

Whether you're doing intent detection, information extraction, semantic role labeling or sentiment analysis, Prodigy provides easy, flexible and powerful annotation options. Active learning keeps you efficient even if your classes are heavily imbalanced.

spaCy

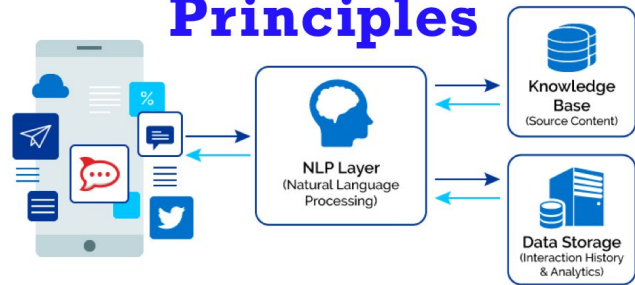


PERSON 1 ORG 2 PRODUCT 3 DATE 4

In a **March 2014** **DATE** interview, **Apple** **ORG** designer **Jonathan Ive** **PERSON** used the **iPhone** **PRODUCT** as an example of **Apple** **ORG**'s ethos of creating high - quality, life - changing products .

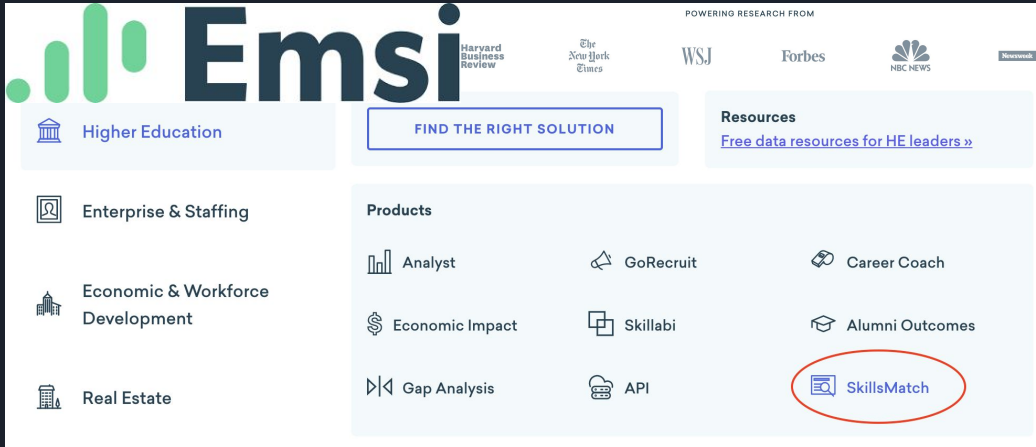


Natural Language Principles



Data Transformation

NLP using Labor Market Analytics & Economic Data API



The image shows the Emsi website header and navigation menu. The Emsi logo is on the left, followed by a navigation bar with links to Higher Education, Enterprise & Staffing, Economic & Workforce Development, and Real Estate. A central button says 'FIND THE RIGHT SOLUTION'. To the right, a 'Resources' section links to 'Free data resources for HE leaders'. Below the navigation bar, a 'Products' section lists various services: Analyst, GoRecruit, Career Coach, Economic Impact, Skillabi, Alumni Outcomes, Gap Analysis, and API. The 'SkillsMatch' product is highlighted with a red circle.

POWERING RESEARCH FROM

Harvard Business Review The New York Times WSJ Forbes NBC NEWS newweek

Higher Education

FIND THE RIGHT SOLUTION

Resources

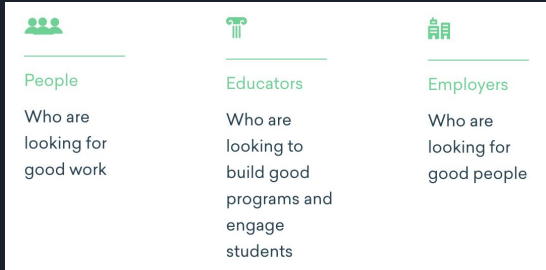
[Free data resources for HE leaders »](#)

Products

Analyst GoRecruit Career Coach

Economic Impact Skillabi Alumni Outcomes

Gap Analysis API SkillsMatch



The image shows the Emsi website footer with three columns: People, Educators, and Employers. Each column has a description of who is looking for good work, programs, and people.

People

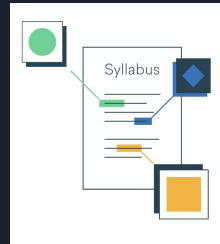
Who are looking for good work

Educators

Who are looking to build good programs and engage students

Employers

Who are looking for good people



OpenSkills API

"30,000+ skills that we've collected from hundreds of millions of job postings, resumes, and online profiles"

What is included with free API access?

- Access to every skill in Emsi's Open Skills Library
- Skill Names, unique machine-readable IDs, and types (technical skill, human skill, certification/license)
- Access to every title in Emsi's Open Titles Library
- Title Names, unique machine-readable IDs, and more
- Autocomplete search
- Limited access to skill extraction

Data Transformation

Emsi Skills API

/versions/{version}/extract

POST Extract skills from document

Client ID: 7k_yuoeqhda9
Secret: vi_RH
Scope: emsi_open



URL Parameters

Name	Description
version	The skills classification version. string Example: latest



Request Body

Property	Description
text string	Document to be used in the skills extraction process
confidenceThreshold number	Filter out skills with a confidence value lower than this threshold <small>Hide details</small> This is an optional attribute. Example: 0.6 Minimum: 0 Maximum: 1 Default: 0.5

We decided on 0.4

httr-package: 'httr' makes http easy.

Response Examples

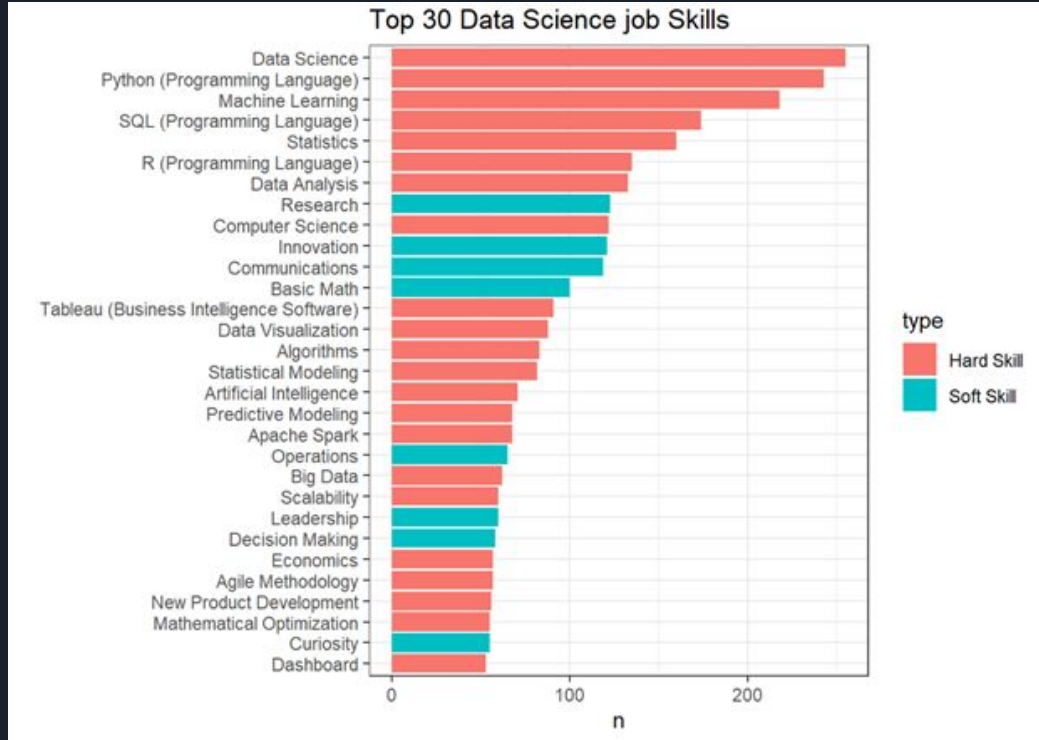
Property	Description
attributions array (objects)	Data attribution information
attributions[].name string	Attribution name
attributions[].text string	Licensing information
data array (objects)	List of extracted skill information
data[].confidence number	A number between 0 and 1 representing the confidence of the skill classification
data[].skill object	Extracted skill information object
data[].skill.type object	Skill type information object
data[].skill.type.id string	Skill type ID
data[].skill.type.name string	Skill type name
data[].skill.id string	Skill ID
data[].skill.name string	Skill name
data[].skill.tags array (objects)	List of tag information of the skill
data[].skill.tags[].key string	Skill tag key
data[].skill.tags[].value string	Skill tag value
data[].skill.infoUrl string	URL for a publicly accessible web page that includes information about the skill

skill	type
Association Rule Learning	Hard Skill
Critical Thinking	Soft Skill

Data Analysis



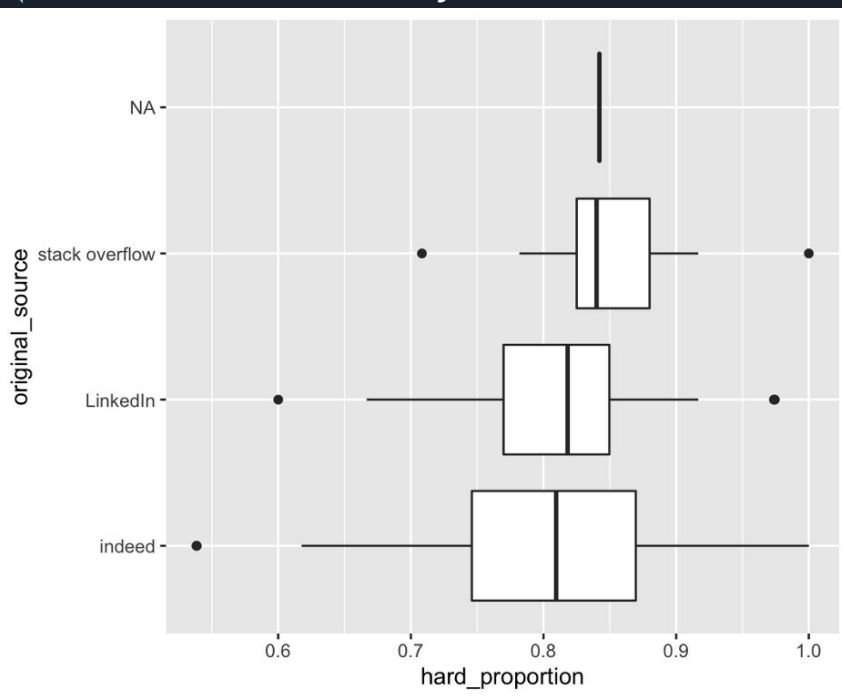
Top 30 Data Science Job Skills



- >900 Unique Skills
- Hard vs Soft Skills
 - $\frac{1}{3}$ of top 12 are soft skills compared to 20% in overall dataset
- Programming skills highly ranked in top 30 skills

Statistical Modeling

“Does Stack Overflow’s distribution of hard_proportion in jobs have a statistically significant difference than the jobs of other sources”



1. Visualize proportional differences
2. Stack Overflow displays a larger proportion of hard skills with a shorter spread
3. Is there a statistical difference?
4. Two Sample T-Test -> p-value = 0.1778
5. Fail to reject the null hypothesis

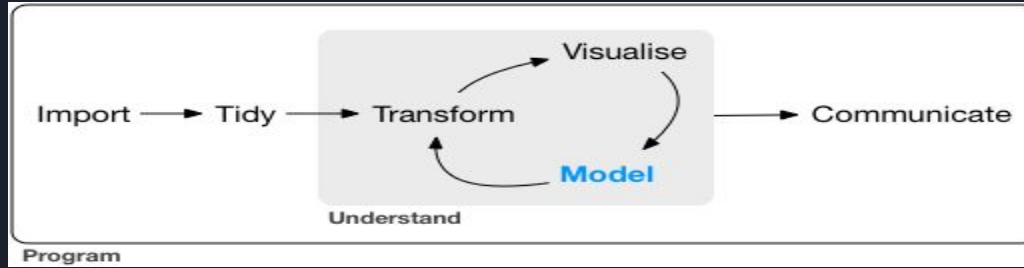
- Two Sample t-test
- data: so_props and other_props
- $t = 1.3536$, $df = 161$, $p\text{-value} = 0.1778$
- 95 percent confidence interval:
 - a. -0.0168604 0.0903353
- sample estimates:
 - a. mean of x mean of y
 - b. 0.8475232 0.8107858

Conclusion



Most Valuable Data Science Skills

Hadley's Data Science Model

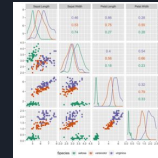


Programming



Import
Tidy
Transform
Visualise

Statistics &
Machine Learning



Model

Communication



Communicate



Next Steps

1. Expand to other jobs / job sources
 - a. POC for Data Science - what do other technology jobs look like?
 - b. Increase sample sizes
2. Analyzing job titles
 - a. We believe we may see different hard to soft skill proportion based on job title (Senior vs Staff Data Scientist)
3. Remove duplicates from sources
 - a. Currently no checks in place to enforce uniqueness
4. Build our own model for skills categorization
 - a. We depend heavily on EMSI to produce results