

# Northeastern University College of Professional Studies

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### ALY 6080: XN Project CoverQuick Individual Draft

Submitted to:

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### About Sponsor

- An Al-based programme called <u>CoverQuick</u> creates unique cover letters and resumes for job applications. By developing a customised cover letter and resume that are suited to the particular job application, CoverQuick hopes to assist job seekers stand out from the competition by utilising the most recent Al techniques.
- One of CoverQuick's distinctive characteristics is its capacity to produce a customised cover letter for each application, ensuring that applicants do not submit generic letters that fall flat with recruiters.
- Products: Prepare Resume, CoverLetter, Track Application, Resume Grader
- Number of Users : 5000 users (Ref till September 2022 )

## EXPLORATORY DATA ANALYSIS (EDA)

#### DATASET PROVIDED

- A. WITH JOB DESCRIPTION
- B. WITHOUT JOB DESCRIPTION

### Research Questions:

- 1. What are the three industries that the majority of CoverQuick's users have applied (With Job Description Dataset)?
- 2. Discover trends in demographics and find which industries yield the best and the worst resumes (CoverQuick provides metrics for defining a "Good" resume).
- 3. Determine the expected age and approximate experience level.
- 4. Determine trends in experience and skills for these target users.

### Planning and Execution

- 1. EDA on job description dataset.
- 2. Dataset splitting for the respective columns which were in json and nested json format.
- 3. Identification and visualization for the top 3 industries that the majority of users have applied.
- 4. Identification and visualization for the approximate age range and experience level.
- 5. Identification and visualization for the trends in experience and skills for these target users.
- 6. Identification and visualization to discover the trends in demographics for the number of candidates registering to the website across globe for resume building.

### **EDA: WITH JOB DESCRIPTION DATASET**

**RAW DATASET** 

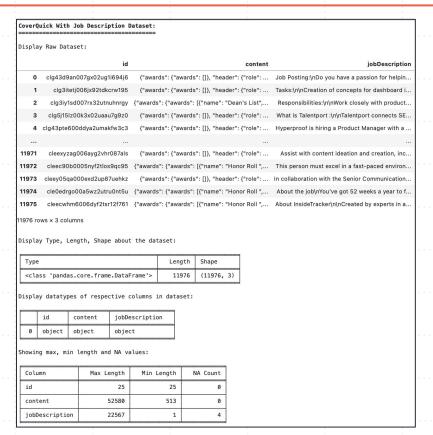
Total Rows: 11976

Total Columns: 3

FINAL DATASET AS OF NOW

Total Rows: 11976

Total Columns: 57



### Distribution of Country Codes w.r.t ID

Country Code was defined using Country Column and after successful analysis we have 0 NULL count which helped us for further visualisation.

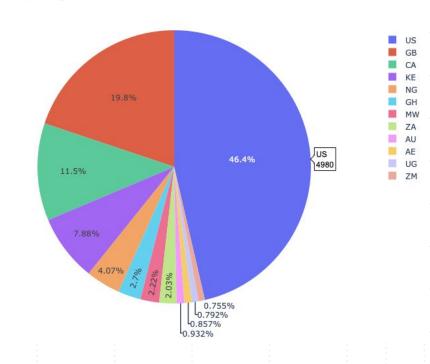
#### COUNTRY\_CODE Non-Null and Null Count:

Total Count Non-Null Count Null Count 0 11976 11976 0

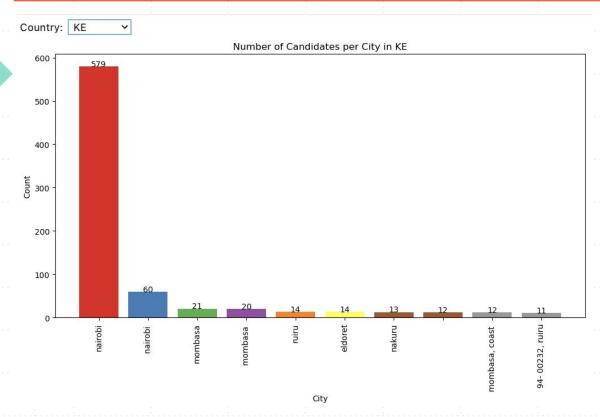
Demographics Explanation: Top 3 countries as per number of users.

- 1. United States has maximum user: 46.4% of total user: 4980 applicants
- 2. Kenya (KE) has 19.8%
- 3. Great Britain (GB): 11.5%

Distribution of Country Codes



### Number of Candidates per City in Kenya

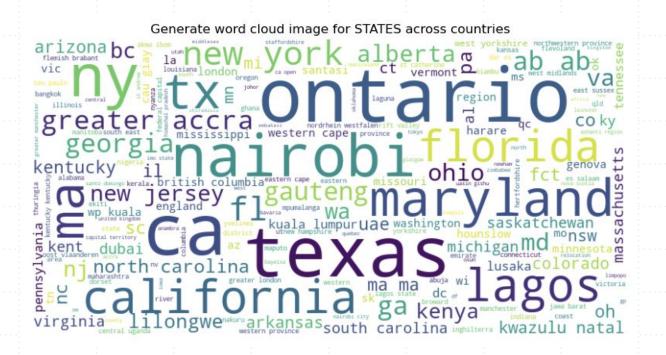


We can determine how many users have registered to CoverQuick's website for resume building by specifying the countries from the drop down.

- 1. Nairobi with 579 users
- 2. Mombasa: 21 users
- 3. Least number of users in Kenya Nakuru, nanyuki cities.: 13, and 12 and 11 respectively

### Generate word cloud image for STATES across countries

- Nairobi, Lagos, Texas, Ontario, Gauteng are the Provinces, States applicants in the worldwide
- Delta states, mefu south, dar es, are the provinces which has less number of applicants



### WordCloud of Most Common Skills and Skills description

#### Common Skills Among the users

- Management
- Technical skills
- Soft Skill (good communication, leadership skills, interpersonal skills)
- Language
- DevOps
- Software

#### Skills as per Description

- 1. Customer Service
- Problem Solving
- 3. Time Management
- 4. Microsoft Office

Competitive Skills required : SQL, Critical thinking, Data Analysis, Analysis

#### WordCloud of Most Common Skills



#### WordCloud of Most Common Skills description

```
communication
Data Entry Social Media
```

### Final Dataset

 After the cleanup, we have 57 columns and their respective data type which will help us to answer our research questions going forward.

	ID	KEYWORDS	SUGGESTEDSKILLS	ROLE CITY	STATE	SUMMARY	ACCOMPLISHMENTS	COUNTRY_CODE	SKILL_DESCRIPTION	PBL_DET/	ILS PBL_PUBLISHE	CRT_NAME	CRT_ISSUER	
0	CLG43D9AN007GX02UG1I694J6	[admissions representative, admissions, uma,	[Compliance, Client, Manages, Interaction, Fin	INDIO	CA	DETAILED AND DRIVEN, I HAVE BUILT STRONG COMMU		US	VERBAL, WRITTEN, AND VISUAL COMMUNICATION, GOA		NaN NaI	QUALIFIED APPLICATOR CERTIFICATE	CALIFORNIA RIVERSIDE AGRICULTURE DEPARTMENT	
1	CLG3ITETJ006JX92TDKCRW195	[dashboard interfaces, lead generation, mark	[Analysis, Collection, Research]	ILMENAU	THURINGIA	DETAILED- ORIENTED UI/UX DESIGNER WITH EXPERIEN		DE	FIGMA, SKETCH, ADOBE XD, FRAMER, MIRO, UXPIN,		NaN NaI	VISUAL ELEMENTS OF USER INTERFACE DESIGN	CALIFORNIA INSTITUTE OF THE ARTS	
2	CLG3IY1SD007RX32UTNUHNRGY	[product, design, development, business req	[Vue, DevOps, Delivery]	PEORIA	ARIZONA	AGILE SOFTWARE ENGINEER WITH 2 YEARS OF EXPERI		us	JIRA, VSC, MYSQL, GIT, BITBUCKET, GITHUB, POS		NaN Nal	I NaN	NaN	
3	CLG5J15LZ00K3X02UAAU7G9Z0	[flexibility, international exposure, dream	0	MALANG		INNOVATIVE DIGITAL MARKETING PROFESSIONAL WITH		GB	MARKETING ANALYTICS, WEBSITE ANALYTICS, PRODUC		NaN NaI	I NaN	NaN	
4	CLG43PTE600DDYA2UMAKFW3C3	[product roadmaps, new features, product enh	[Curiosity]	CALGARY	АВ	PASSIONATE JOB SEEKER WITH STRONG ORGANIZATION		CA	CRITICAL AND ANALYTICAL THINKING, TIME MANAGEM	***	NaN Nal	SCRUM MASTER CERTIFICATION	LEARN QUEST	
11971	CLEEXYZAG006AYG2VHR087ALS	[content, ideation, creation, camera, cont	[Instagram, Calendar, TikTok]	BROOKLYN	NY	DEPENDABLE VIDEOGRAPHER AND VIDEO EDITOR WITH		US	PROJECT MANAGEMENT, SELF- DRIVEN, PRODUCTION PL	***	NaN Nal	I NaN	NaN	
11972	CLEEC90B0005NYF2TLOS9QC95	[adobe premiere, adobe after effects, adobe 	[Broadcast, Promotional, Broadcast & Promotion	PEABODY	МА			US	FACEBOOK LIVE, TWITCH, OBS, XSPLIT		NaN Nal	LEARN HTML COURSE	CODECADEMY	
11973	CLEEY05QA000EXD2UP87UEHKZ	[write, content, graphics, imagery, social	D	CRIVITZ	WI	PERSONABLE AND HARDWORKING PROFESSIONAL WITH E		GB	NaN		NaN Nal	INTERNATIONAL ORGANIZATION MANAGEMENT	UNIVERSITY OF GENEVA	
11974	CLE0EDRGO00A5WZ2UTRU0NT5U	[accountable, challenges, social media, bes	О	PEABODY	МА	PROFESSIONAL WITH OVER A DECADE OF EXPERIENCE		us	FACEBOOK LIVE, TWITCH, OBS, XSPLIT		NaN Nal	LEARN HTML COURSE	CODECADEMY	
11975	CLEECWHM6006DYF2TSR12F761	[product, product team, product managers, s	[Communicate]	PEABODY	МА			us	FACEBOOK LIVE, TWITCH, OBS, XSPLIT		NaN NaI	LEARN HTML COURSE	CODECADEMY	
11976 ro	ws × 57 columns													

df.dtypes	
ID	object
KEYWORDS	object
SUGGESTEDSKILLS	object
ROLE	object
CITY	object
STATE	object
SUMMARY	object
ACCOMPLISHMENTS	object
COUNTRY_CODE	object
SKILL_DESCRIPTION	object
SKILL	object
EDU GPA	object
EDU_MINOR	object
EDU_AWARDS	
EDU_SCHOOL	object
	object
EDU_PROGRAM	object
EDU_LOCATION	object
EDU_COURSEWORK	object
	datetime64[ns]
EDU_GRAD_YEAR	int64
BIRTH_YEAR	int64
AGE_RANGE	int64
VLNTR_TITLE	object
VLNTR_ENDDATE	datetime64[ns]
VLNTR_LOCATION	object
VLNTR_STARTDATE	datetime64[ns]
VLNTR_DESCRIPTION	object
VLNTR_ORGANIZATION	object
EXP_TITLE	object
EXP_COMPANY EXP_ENDDATE EXP_LOCATION EXP_STARTDATE	object
EXP_ENDDATE	datetime64[ns]
EXP_LOCATION	object
EXP_STARTDATE	datetime64[ns]
EXP_DESCRIPTION	object
PRJ_LINK	object
PRJ_TITLE	object
PRJ_SKILLS	object
PRJ_ENDDATE	datetime64[ns]
PRJ_STARTDATE	datetime64[ns]
PRJ_DESCRIPTION	object
REF_NAME	object
REF_EMAIL	object
REF_PHONENUMBER REF_RELATIONSHIP	object
REF_RELATIONSHIP	object
PBL_DATE	datetime64[ns]
PBL_LINK	object
PBL_NAME	object
PBL_DETAILS	object
PBL_PUBLISHER	object
CRT_NAME	object
CRT_ISSUER	object
CRT_DATERECEIVED	datetime64[ns]
AWD_NAME	object
AWD_ISSUER	object
AWD_DETAILS	object
AWD_DATERECEIVED	datetime64[ns]
AWD DESCRIPTION	object
dtunes object	,

df.dtvpes

### Determine the approximate age range and experience level

• Predefined duration thresholds are established in order to classify experience levels. These levels are classified as 'BEGINNER' for experience durations up to 1 year (365 days), 'INTERMEDIATE' for durations ranging from 1 to 2 years (365 to 730 days), and 'ADVANCED' for durations between 2 to 3 years (730 to 1095 days).

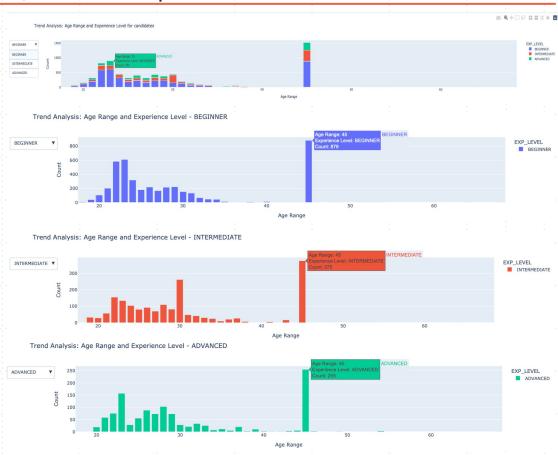
 Subsequently, the DataFrame is modified to incorporate an additional column called 'EXP\_DURATION', which denotes the duration of experience calculated in days. Additionally, an 'EXP\_LEVEL' column is introduced, which classifies the experience level based on predefined

thresholds.

	ID	COUNTRY_CODE	BIRTH_YEAR	AGE_RANGE	EXP_DURATION	EXP_LEVE
0	CLG43D9AN007GX02UG1I694J6	US	1997	26	NaN	NaN
1	CLG3ITETJ006JX92TDKCRW195	DE	1995	28	699.0	INTERMEDIATE
2	CLG3IY1SD007RX32UTNUHNRGY	US	1998	25	672.0	INTERMEDIAT
3	CLG5J15LZ00K3X02UAAU7G9Z0	GB	1998	25	122.0	BEGINNE
4	CLG43PTE600DDYA2UMAKFW3C3	CA	1999	24	92.0	BEGINNE
					•••	
1971	CLEEXYZAG006AYG2VHR087ALS	US	1987	36	519.0	INTERMEDIAT
1972	CLEEC90B0005NYF2TLOS9QC95	US	1998	25	NaN	Nai
1973	CLEEY05QA000EXD2UP87UEHKZ	GB	1993	30	730.0	INTERMEDIAT
1974	CLE0EDRGO00A5WZ2UTRU0NT5U	US	1998	25	NaN	Nai
1975	CLEECWHM6006DYF2TSR12F761	US	1998	25	NaN	Nai

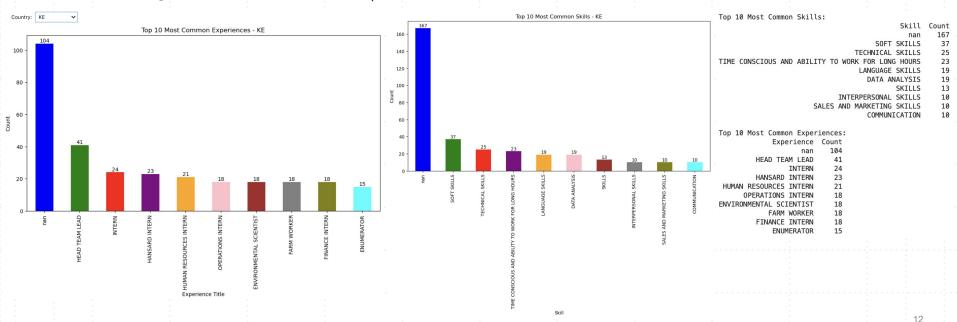
### Determine the expected age and experience level

- The x-axis represents the age ranges, ranging from 18-24, 25-34, 35-44, 45-54, and 55+. The y-axis represents the count of candidates. Each bar in the chart is segmented into different colors representing different experience levels.
- By selecting different experience levels from the dropdown menu, the chart dynamically updates to show the trend analysis specifically for that experience level. The title of the chart also changes accordingly to provide focused insights.
- Hovering over each bar provides additional information, including the specific age range, experience level, and the corresponding count of candidates.
- The visualization aims to provide a clear and visually appealing representation of the distribution of candidates across different age ranges and experience levels, allowing for a quick and comprehensive understanding of the trends in your project data.



### Determine trends in experience and skills for the target users.

A bar chart can be utilized to visually represent the top 10 prevalent experiences and skills. By utilizing a dropdown menu, users have the option to choose specific countries and observe the corresponding top 10 experiences and skills associated with those countries. This graphical representation allows for a clear understanding of the most common experiences and skills across different countries.



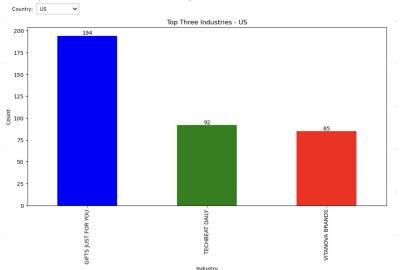
### Top three industries that the majority of CoverQuick's users have applied:

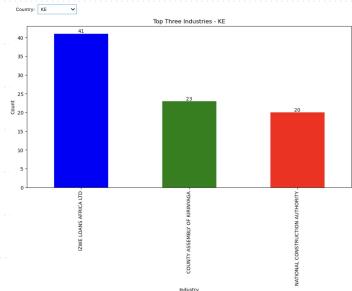
As shown below, from the total dataset with job description, we can see that the top three industries/company candidates apply to are mostly GIFTS JUST FOR YOU (194), LISAP (163), and **EDMONTON FIRE RESCUE (159).** 

Top Three Industries: GIFTS JUST FOR YOU LIVINGSTONIA SYNOD AIDS PROGRAMME (LISAP) EDMONTON FIRE RESCUE

We can also select the countries from the dropdown and get the

top three industries per country wise.





194

163

159

## Discover trends in demographics and find which industries yield the best and the worst resumes.

#### Resume Optimality Criteria

- 1. Important Sections: This may include and not be limited to: work experience, education, projects, as the most important and relevant sections.
- 2. Resume Length: The solid resume length may be between 300-500 words, however; if the length is outside this range, it may not mean a resume is poor.
- 3. Use of action verbs: Direct use of action verbs in the bullets of a resume will ensure a resume will perform better.
  - 4. No use of pronouns: Resumes should not contain pronouns such as I, we or me written in the document.
  - 5. Excessive bullet points: A resume experience or section should not have an excessive number of bullet points. If a section has over 10 bullet points, it is looked upon unfavourably.
- ✓ 6. Spelling Mistakes: A resume with spelling errors is immediately penalized against.
  - 7. Excessive sentence or bullet length

#### RESUME LENGTH

EXP_DURATION	EXP_LEVEL	RES_LEN
nan	NaN	188
699.0	INTERMEDIATE	259
672.0	INTERMEDIATE	188
122.0	BEGINNER	161
92.0	BEGINNER	136
519.0	INTERMEDIATE	145
nan	NaN	484
730.0	INTERMEDIATE	155
nan	NaN	349
nan	NaN	451

We iterate over the specified columns and count the total number of words. The word count is then added as a new column 'RES\_LEN' to the DataFrame.

Finally, we will map it to 'POOR' if it's below 300 and 'GOOD' if it's greater than or equal to 300 using our score-card.

#### RES\_LEN Non-Null and Null Count:

Total Count Non-Null Count Null Count 0 11976 11976 0

### **ACTION VERBS**

EXP_DURATION	EXP_LEVEL	RES_LEN	ACTN_VERB
nan	NaN	188	47
699.0	INTERMEDIATE	259	58
672.0	INTERMEDIATE	188	56
122.0	BEGINNER	161	49
92.0	BEGINNER	136	41
519.0	INTERMEDIATE	145	40
nan	NaN	484	116
730.0	INTERMEDIATE	155	36
nan	NaN	349	89
nan	NaN	451	109

```
import nltk
from nltk import word_tokenize
from nltk.corpus import stopwords
from nltk.corpus import wordnet

# Download necessary NLTK resources if not already downloaded
nltk.download('punkt')
nltk.download('stopwords')
nltk.download('wordnet')
nltk.download('omw-1.4')
```

### ACTN\_VERB Non-Null and Null Count:

Total Count Non-Null Count Null Count 0 11976 11976 0

### **SPELLING MISTAKES**

EXP_DURATION	EXP_LEVEL	RES_LEN	ACTN_VERB	SPLNG_MST
nan	NaN	188	47	ŕ
699.0	INTERMEDIATE	259	58	
672.0	INTERMEDIATE	188	56	ś
122.0	BEGINNER	161	49	1
92.0	BEGINNER	136	41	1
519.0	INTERMEDIATE	145	40	1
nan	NaN	484	116	1
730.0	INTERMEDIATE	155	36	1
nan	NaN	349	89	1
nan	NaN	451	109	,

```
'aardvark'
 'ab'
 'aback'
 'abacus'
 'abalone'
 'abandon'
 'abandoned'
 'abandoning'
 'abandonment'
'abandons'
 'abase'
'abased'
 'abate'
 'abated'
 'abatement'
 'abates'
 'abattoir'
 'abba'
 'abbas'
'abbess'
 'abbey'
 "abbey's"
 'abbeys'
 'abbie'
'abbies'
'abbot'
 "abbot's"
'abbots'
 'abbott'
 "abbott's"
 'abbreviate'
 'abbreviated'
 'abbreviation'
 'abbreviations
 'abby'
 "abby's"
 'abc'
'abdal'
'abdicate'
 'abdicated'
'abdicating'
 'abdication'
 'abdomen'
"abdomen's"
 'abdomens'
 'abdominal'
 'abdominals'
 'abduct'
 'abducted'
Output of this cell has been trimmed on the initial display.
```

Displaying the first 50 top outputs.

Click on this message to get the complete output.

```
!pip install spellchecker
import pandas as pd
from spellchecker import SpellChecker
splng_mstk.SPLNG_MSTK.unique()
array([ 1, 0, 2, 24, 9, 3,
SPLNG_MSTK Non-Null and Null Count:
               Non-Null Count
   Total Count
         11976
                         11976
```

Null Count

### **IMPORTANT SECTIONS**

E	XP_DURATION	EXP_LEVEL	RES_LEN	ACTN_VERB	SPLNG_MSTK	IMP_SEC	
	nan	NaN	188	47	1	1	
	699.0	INTERMEDIATE	259	58	1	1	
	672.0	INTERMEDIATE	188	56	Ī	1	
	122.0	BEGINNER	161	49	1	1	
	92.0	BEGINNER	136	41	1	1	
	519.0	INTERMEDIATE	145	40	1	1	
	nan	NaN	484	116	1	1	
	730.0	INTERMEDIATE	155	36	1	1	
	nan	NaN	349	89	1	1	
	nan	NaN	451	109	1	1	

We included and not be limited to: work experience, education

We are checking both the conditions and mapping value to 1 in IMP\_SEC

imp\_sec.loc[~imp\_sec['EXP\_DURATION'].isna() &
(imp\_sec['EDU\_GRAD\_YEAR'] != 1900), 'IMP\_SEC'] = 1

1 8486 0 3490

### NEXT ACTION, FUTURE ACTION AND LIMITATION

#### We will be checking

- No use of Pronouns
- Excessive Bullet length -> Penalise it using our score card generator

#### SCORE CARD:

We will be considering all the values in the respective columns that we have created and based on the the final column "SCORECARD" we will be defining if the resume is good or bad resume.

#### **FUTURE SCOPE:**

Based on our scorecard analysis and considering relevant columns for Suggested Key Skills and Market analysis for respective Jobs which the candidate will be applying for, we will be suggesting and modifying the resume and showcase the analysis for better understanding to the user.

#### LIMITATIONS

- No control over the database.
- No overview on the current code base.
- Higher computing power.
- Additional services which can be paid.



## Thank You!

```
1 def gratitude():
2 print("Thank you.")
3
```