Analyzing Patent Data

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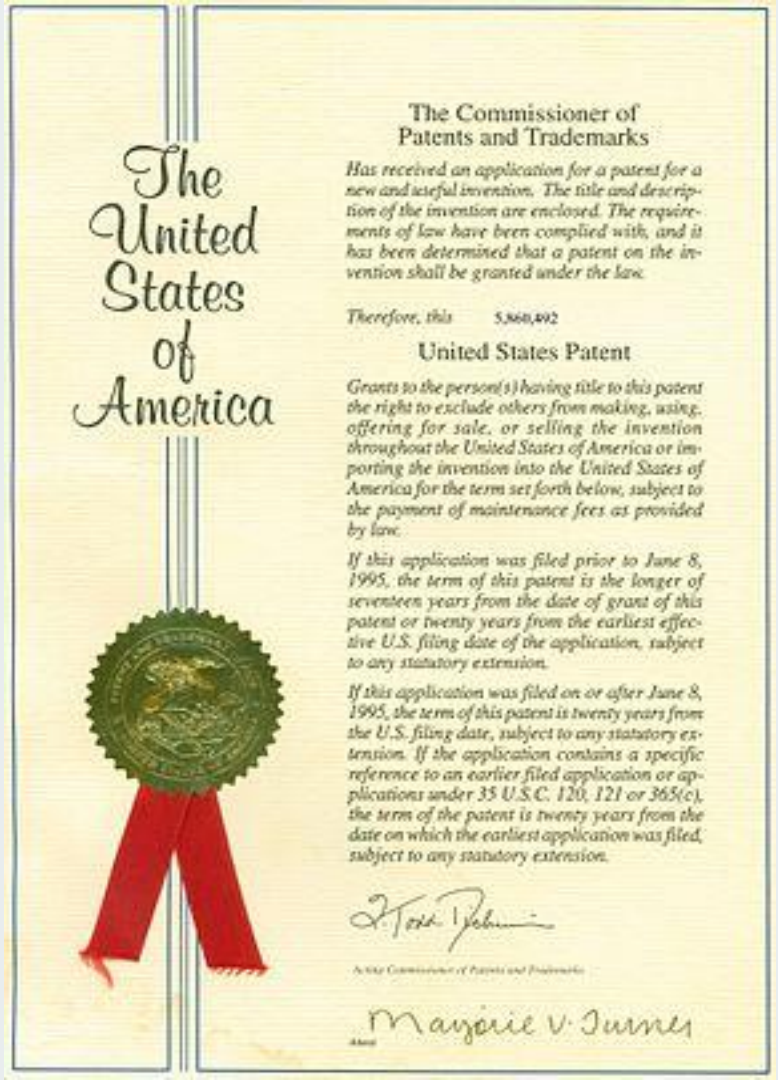
Mandeep Kaur



What is Patent?

- A legal protection which gives an inventor the right to exclude others from performing certain activity in the country of issuance.
- Sanctioned monopoly for a set number of years in exchange for disclosure to the public.
- Does not give the inventor the right to make, use or sell the patented invention

Problem Statement

Our project aims at analyzing patent data and getting useful insights out of it such as,

- Top Potential Competitors
- Technology in Trend (2018)
- Similar Patents

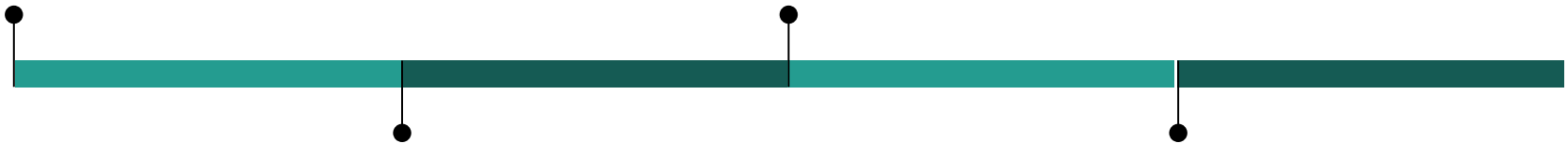
Approach

Data Collection

Google Public Patent Data

Feature Extraction

TF-IDF Matrix



Data Preparation

Data Cleaning
Removal of stop words

Model Training

K-means Clustering
Random Forest Classification

Google Patent Public Dataset (2018)

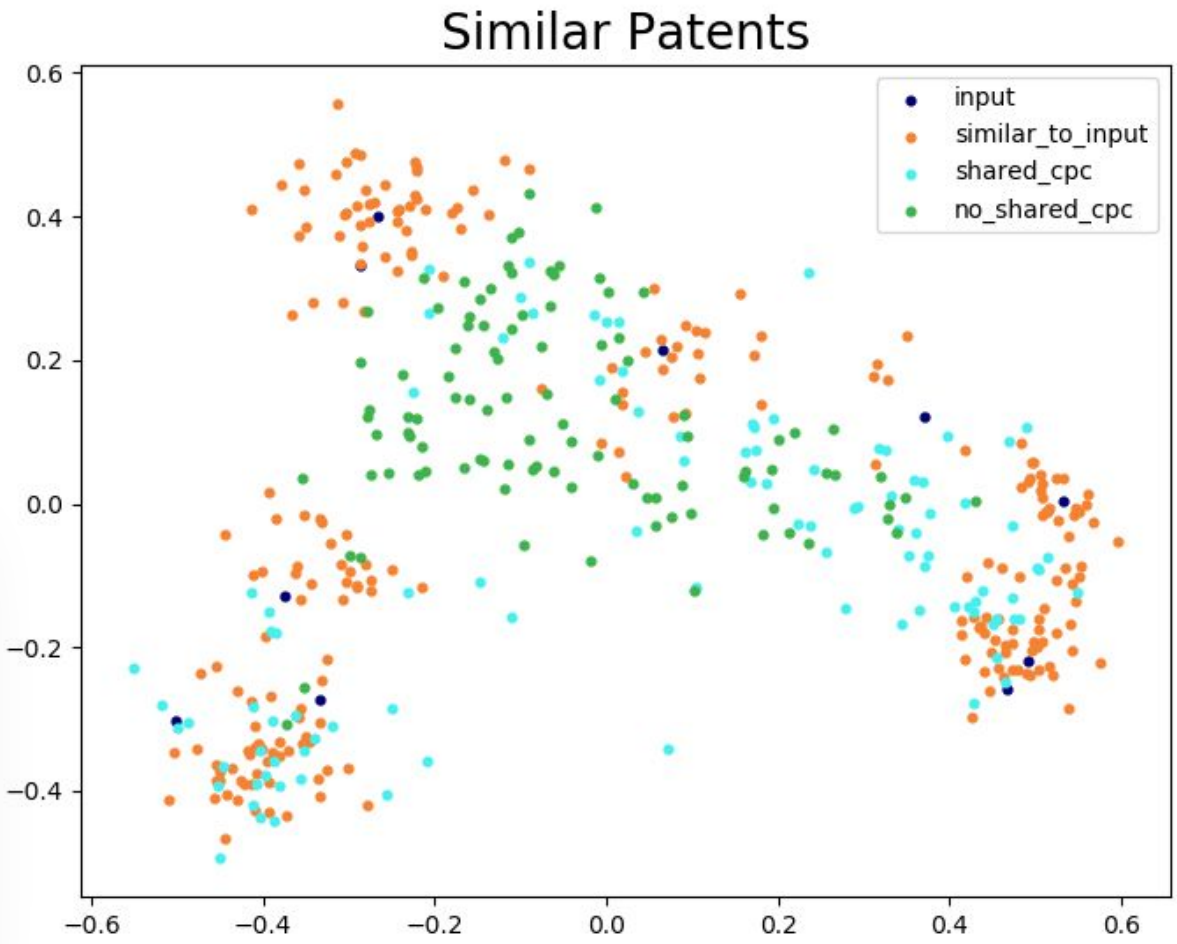
- Data Collected using BigQuery from Google cloud platform
- 1000 samples and 9 features

	publication_number	country_code	title_localized	abstract_localized	description_localized	publication_date	inventor	assignee	cpc
2	US-1705778-A	US	[{'text': 'Sound-abs	[]	[{'text': 'March 19, 1929. T. B. MU	19290319	['MUNROE TREAD	['Munroe']	[{'code': 'Y10S454/906', 'inventiv
3	US-1714689-A	US	[{'text': 'Oscillation	[]	[{'text': 'y 23, 1929- .J. A. MILLEF	19290528	['MILLER JAMES A	['John Flam', 'I	[{'code': 'H03B11/08', 'inventive'
4	US-1782733-A	US	[{'text': 'Bag-filling r	[]	[{'text': 'Nov. 25, 1930. s. H. LILL	19301125	['LILLY SCOTT H.'	['Scott H Lilly']	[{'code': 'B65B1/18', 'inventive':
5	US-1843645-A	US	[{'text': 'Discharge t	[]	[{'text': 'Feb. 12,1932. MEYER ET	19320202	['MEYER FRIEDRI	['Electrons Inc'	[{'code': 'Y10S315/01', 'inventive
6	US-1898068-A	US	[{'text': 'Welding ele	[]	[{'text': 'Feb. 21, 1933. L. B. THO	19330221	['THOMPSON LUC	['Gen Electric']	[{'code': 'Y10T428/12132', 'inver
7	US-1911286-A	US	[{'text': 'Push buttor	[]	[{'text': 'May 3o, 1933. E. PALMIE	19330530	['PALMIERI EMAN	['Palmieri Emar	[{'code': 'F16K21/14', 'inventive':
8	US-1923363-A	US	[{'text': 'Container',	[]	[{'text': '1933- A. E. FREDRICKSC	19330822	['FREDRICKSON A	['Axel E Fredric	[{'code': 'F25D31/002', 'inventive
9	US-1929859-A	US	[{'text': 'Photo-elect	[]	[{'text': 'J. B. s'rRAuss 1,929	19331010	['STRAUSS JOSEI	['Joseph B Stra	[{'code': 'B61L29/24', 'inventive':
10	US-1936524-A	US	[{'text': 'Method and	[]	[{'text': 'Nov. 21, 1933. A. PLACE	19331121	['PLACEK ADOLPI	['Placek Adolp	[{'code': 'B01J2219/1944', 'inver
11	US-1984692-A	US	[{'text': 'Door opera	[]	[{'text': 'Dec. 18, 1934 NICHQLs	19341218	['NICHOLS FRED	['Fred L Nichol	[{'code': 'E05F15/53', 'inventive':
12	US-1985636-A	US	[{'text': 'Refrigeratio	[]	[{'text': 'Dec. 25, 1934 B, 5 055 1	19341225	['FOSS BENJAMIN	['B F Sturtevan	[{'code': 'Y10S204/06', 'inventive
13	US-1989786-A	US	[{'text': 'Base and b	[]	[{'text': 'Feb. 5, 1935.- E. c. BRUI	19350205	['BRUECKMANN E	['Westinghouse	[{'code': 'H01J5/54', 'inventive':
14	US-1989925-A	US	[{'text': 'Process of	[]	[{'text': 'Patented F eh. 5, 1935 G	19350205	['HOOVER GEORG	['American Rol	[{'code': 'Y10T428/12618', 'inver
15	US-2001040298-A1	US	[{'text': 'Method of r	[{'text': 'A wafer rec	[{'text': 'BACKGROUND OF THE	20011115	['BABA SHUNJI', '	['Shunji Baba',	[{'code': 'H01L2221/68377', 'inve
16	US-2002037896-A1	US	[{'text': 'Bicyclic cor	[{'text': 'Substituted	[{'text': 'FIELD OF THE INVENTIC	20020328	['BOGENSTAETTE	['Michael Boge	[{'code': 'C07D295/205', 'inventiv

Descriptive Analysis of Data

- We found similar patents by comparing their CPC codes using BigQuery.
- Data Visualization using PCA

CPC - Cooperative Patent Classification
It has 9 classes & 250000 sub-classifications:
A- Human Necessities
B- Performing Operations;Transporting
C-Chemistry;Metallurgy
D-Textiles;Paper
E-Fixed Construction
F-Mechanical Engineering
G-Physics
H-Electricity
Y-General tagging of new Technology

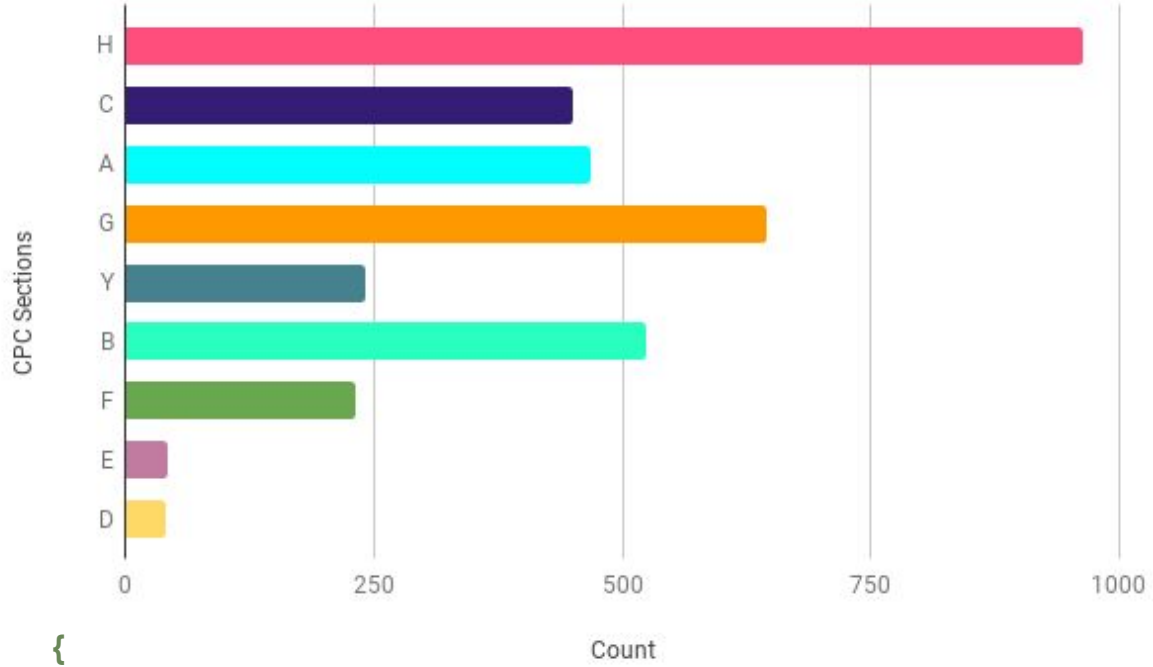


Descriptive Analysis of Data

CPC Codes

A 61 k 39/00 11

- Section: Human Necessities
- Class: 2 digit No: Medical/vertinity
- Sub-Class (A-Z): preparations for medical etc.
- Main Group: antigens or antibodies in prep
- Sub-Group: Cancer antigens




```
{
    "A": "Human Necessities",
    "B": "Operations and Transport",
    "C": "Chemistry and Metallurgy",
    "D": "Textiles",
    "E": "Fixed Constructions",
    "F": "Mechanical Engineering",
    "G": "Physics",
    "H": "Electricity",
    "Y": "Emerging Cross-Sectional Technologies"
}
```

Data Preparation

Data Cleaning → Removal of Punctuations and Stop Words → Tokenization

```
[{'text': 'A high-speed, soft-recovery
semiconductor device that reduces leakage
current by increasing the Schottky ratio
of Schottky contacts to pn junctions',
'language': 'en'}]
```

```
[high speed soft-recovery
semiconductor device reduces
leakage current increasing
Schottky ratio Schottky
contacts pn junctions]
```

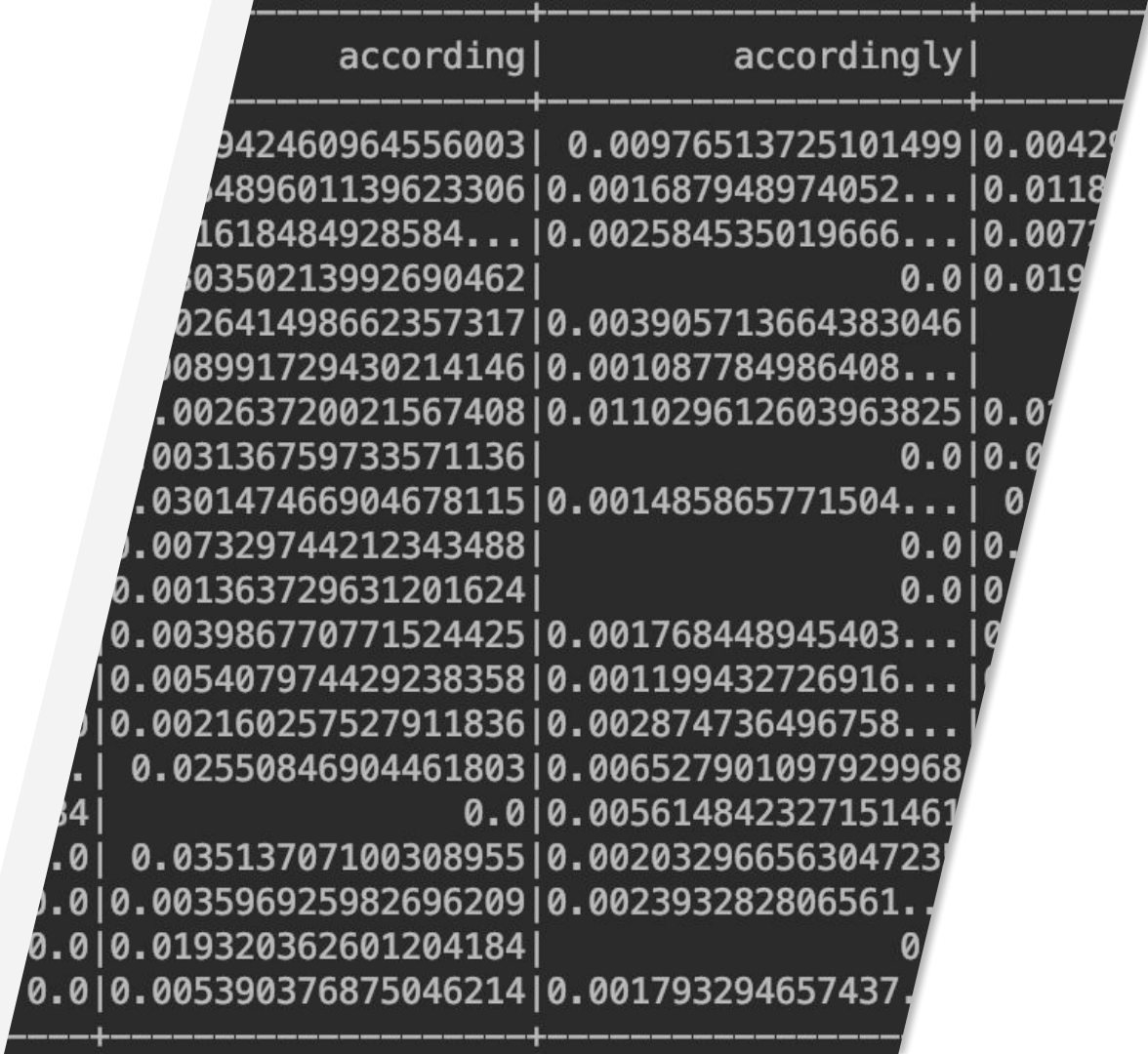
```
["high" "speed" "soft" "recovery"
"semiconductor" "device" "reduces"
"leakage" "current" "increasing"
"Schottky" "ratio" "Schottky"
"contacts" "pn" "junctions"]
```

Feature Extraction

Tf-IDf matrix

Parameters used:

- min_df=0.4
- smooth_idf=True
- lowercase=True
- analyzer='word'
- use_idf=True



Term Frequency -Inverse Document Frequency (TF-IDF)

pub num	accompanying	accordance	according	accordingly	addition	additional	advantages	
US-2001040298-A1	0.001009510837231...	0.0	0.013942460964556003	0.00976513725101499	0.004296338126151883	0.0	0.001047929056912...	0
US-2002037896-A1	0.0	0.001717422991359...	0.016489601139623306	0.001687948974052...	0.011882269499434035	0.016383482828164146	0.005434180998261235	
US-2002055159-A1	0.0	8.765549063751532E-4	0.001618484928584...	0.002584535019666...	0.007201697395568372	0.003251873310624...	4.622582827217685E-4	8
US-2002095050-A1	0.0	0.0	0.030350213992690462	0.0	0.019384871906886703	0.0	0.0	
US-2003052383-A1	0.002018845291143...	0.0	0.02641498662357317	0.003905713664383046	0.0	0.0	0.012574050108093543	0
US-2003052813-A1	0.001124542035742...	0.0	0.008991729430214146	0.001087784986408...	0.0	0.0	0.0	
US-2003099932-A1	0.0	5.10100250309864E-4	0.00263720021567408	0.011029612603963825	0.015440299483161524	0.012435688099987787	5.380109426328707E-4	0
US-2003118999-A1	0.0	0.0	0.003136759733571136	0.0	0.007346049709112546	0.013505150596561058	0.004479476892686568	
US-2004038943-A1	0.0	0.0	0.030147466904678115	0.001485865771504...	0.01307463807289475	0.004807344441215494	0.0	
US-2006189580-A1	0.0	0.0	0.007329744212343488	0.0	0.015019984370287561	0.003944727514612...	0.001308413084571...	
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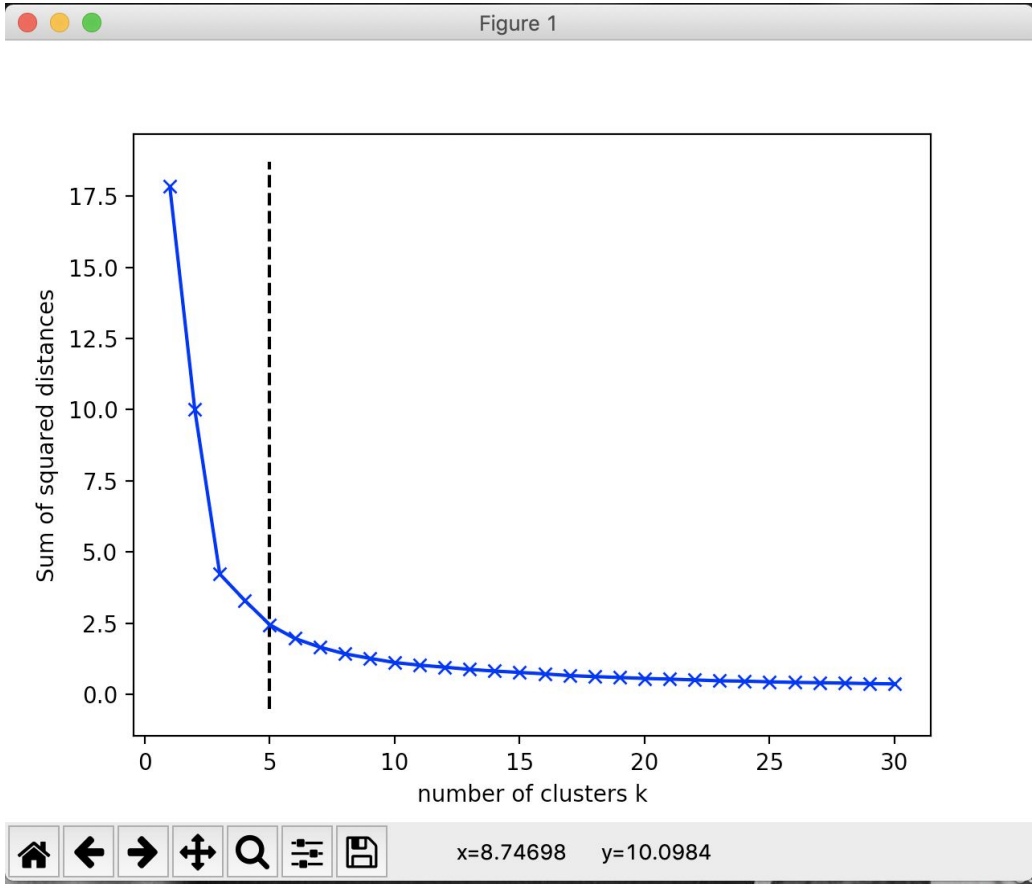
K means Clustering

- Unsupervised Learning algorithm.
- Elbow point found at K = 5 (represented by dashed line)



Principal Component Analysis (PCA)

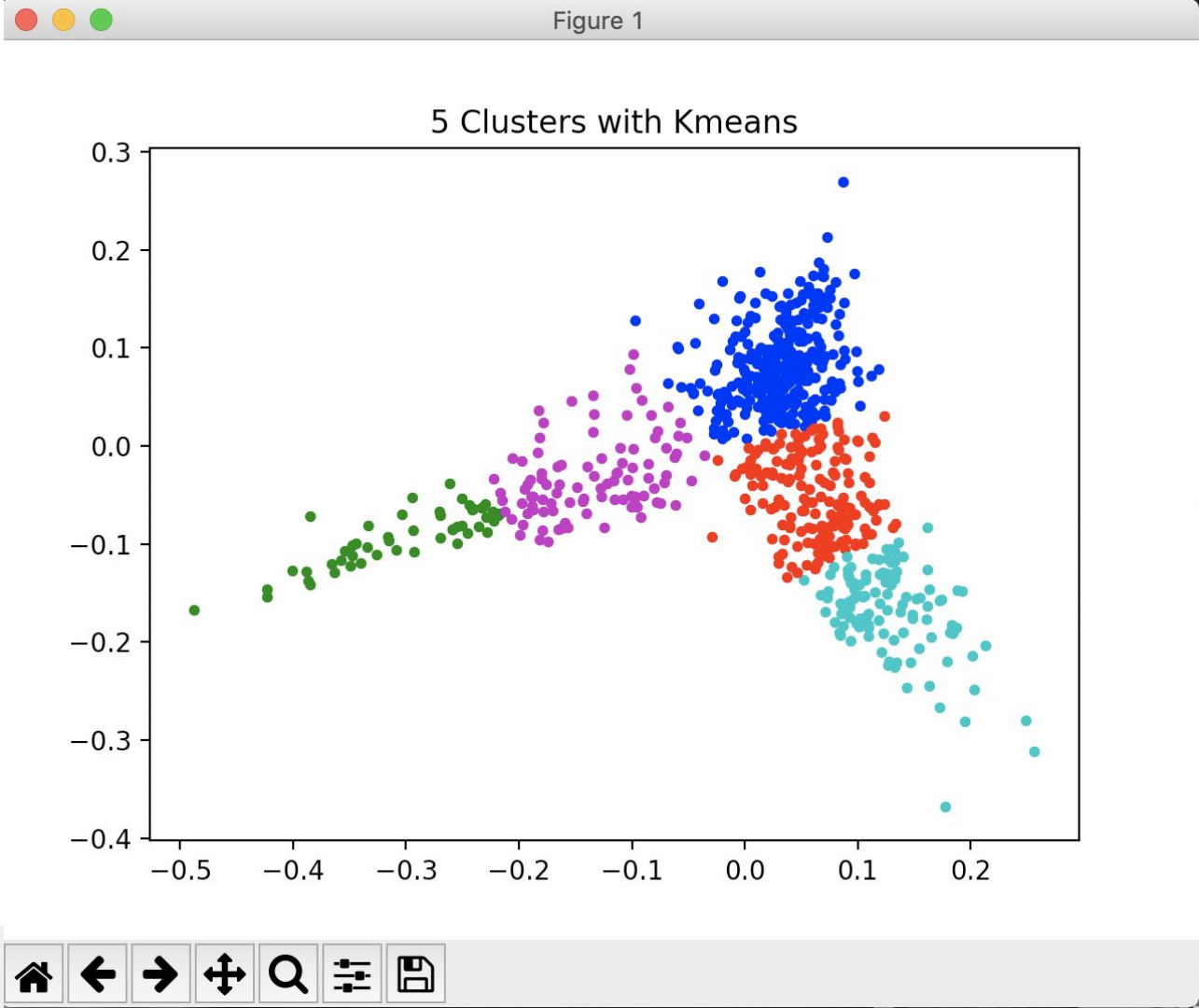
The main idea of **principal component analysis**(**PCA**) is to reduce the dimensionality of a **data** set consisting of many variables correlated with each other, either heavily or lightly, while retaining the variation present in the dataset, up to the maximum extent.

IN BRIEF, PCA IS USED TO :

- **Reduce number of dimensions in data**
- **Find patterns in high-dimensional data**
- **Visualize data of high dimensionality**

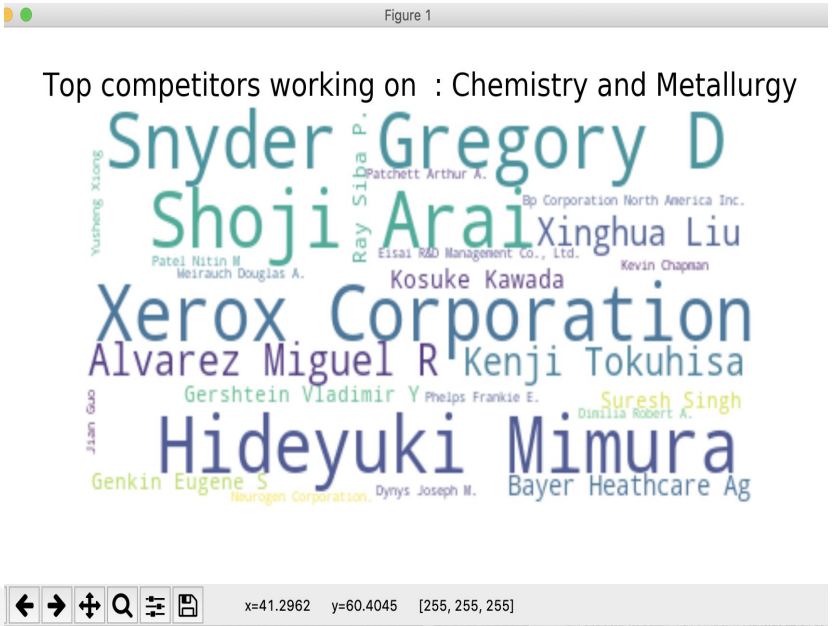
High dimensional data was converted to 2-D for better visualization of clusters using PCA.

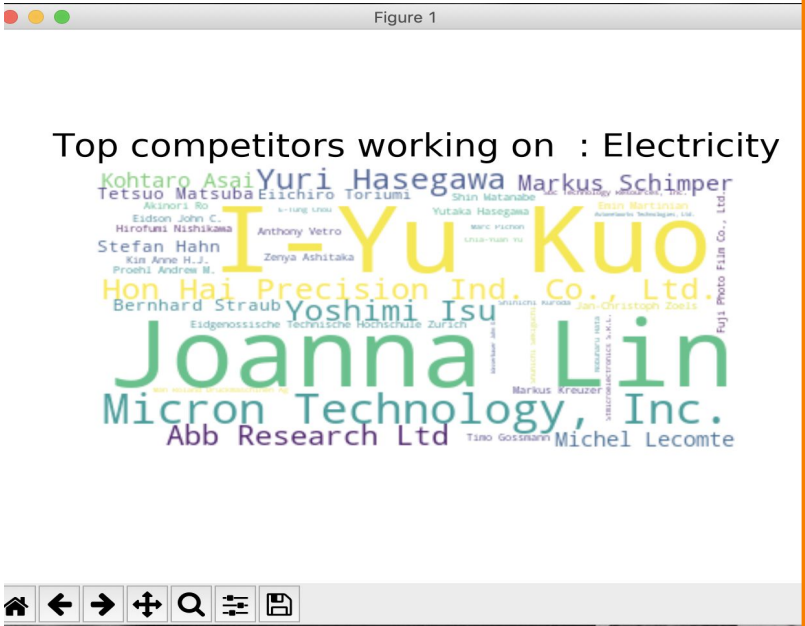
- **Silhouette**, with squared euclidean distance = 0.054



Analysis

- Word clouds generated for two random clusters representing top potential competitors clustered together.
- Technologies in trend was found to be Electricity, Chemistry and Metallurgy which maximum count n the clusters.





Random Forest Classifier

- Supervised Learning Algorithm
- Training Data - K clusters of patents were taken and labelled.
- 10 fold cross validation
- Accuracy obtained is 81%

70% Training Data

30% Testing Data

```
Test Error = 0.182573
accuracy 0.8174273858921162
RandomForestClassificationModel (uid=RandomForestClassifier_3dd733255c57) with 100 trees
```

Conclusion

- We are able to find the potential competitors, which can benefit a company to knowing its competitor.
- By analysing similar cpc codes (matching the first character of the cpc code) in the cluster, we can predict in which cluster does a new patent belongs to using classification.

Challenges

- **Cleaning the data:**

  Some of the data had inconsistent type for a column . For Example , in some rows the assignee names were in quotes and in the some of the rows the assignee data was in string representation of a dictionary

- **BigQuery quota exceeded error** :

  BigQuery gave quota exceeded error while writing custom queries.

Future work and Improvements

1. Research is needed to extract better meaningful words from the patent description, title and abstract and neglect non pertinent words.
2. More features can be extracted from the data, dates and countries.
3. The results are restricted to 1000 patents. We would like to do the analysis on complete patent dataset using multiple nodes which would yield more accurate results.

Thank you!