

# Exploring 6 with Data

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# 360Giving & their Brief

#### 360Giving's vision

for UK grantmaking to be more informed, effective & strategic

#### Our project brief

better understand data in the 360 giving database 'Grant Nav'

#### **Motivation**

- limited data literacy amongst some publishers, funders, grant recipients
- use skilled data analysis techniques to gain new insights
- intended audience 360Giving, funders & recipients

348K grants 198K recipients 115 **funders** 











# Case Study on The Co-operative Group "The Co-op"

Co-operative Group between 2017 and 2018



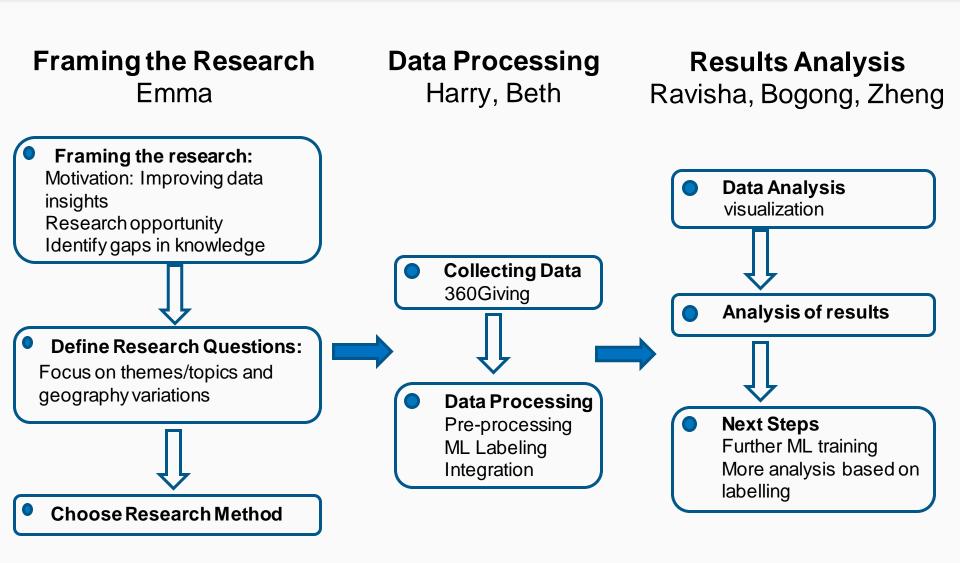
### Co-op funding

- 1% of products and service bought by members are allocated to local communities
- 4,000 local causes to choose from



From the 360Giving data registry
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https://insights.threesixtygiving.org/file/a002400000gzkbKAAQ

# Research Strategy & Project Plan



# **Research Questions**

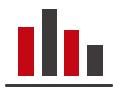
What themes of grants are awarded by The Co-op?



 Where are the local causes located? Is there correlation between themes and locations?



Is there any geographical variation in the amount awarded to recipients based on theme?

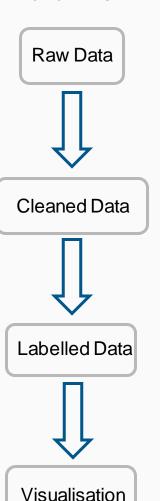


# **Data Processing**

# **Tools** R packages: **PostcodeioR** tidyverse Google Cloud Natural Language API & Python Power BI

Power BI

#### **Data Flow**



#### **Process**

- Data Cleansing & Feature Selection
  - Remove redundant columns/rows
  - Convert postcodes to region, county, country and latitude & longitude data

#### Labelling the Data

- Labelling the data allows us to categorize the grants into themes, for further analysis
- Python code to run the data to Google's API

# **Labelling/Categorization of Grants**

Part 1



Concatenate title, description and recipient name columns of each grant

4	Α	В	<u> </u>	D				
1		ïldentifier	Title	Description				
2	1	360G-COOPGROUP-11483	Grant to Friends of Castle Park Primary School	We want to purchase a secure outdoor classroom for the scho				
3	2	360G-COOPGROUP-11488	Grant to Friends of West Wimbledon	Funds will help us develop the playground area to include a se				
4	3	360G-COOPGROUP-11490	Grant to Laira Greenshoots Community Forum	We want to buy new playground equipment and playground p				
5	4	360G-COOPGROUP-11494	Grant to High Five	We want to provide social activities for disabled children and t				
6	5	360G-COOPGROUP-11501	Grant to Rose Green Infant School PTA	We want to replace old and tired white boards with new inter-				
7	6	360G-COOPGROUP-11503	Grant to Children and Families in Grief Ltd	We want to give bereavement counselling and support to child				
8	7	360G-COOPGROUP-11507	Grant to 1st Aberchirder guides	We would like take the girls on a trip to PGL Adventure Centre				
9	8	360G-COOPGROUP-11510	Grant to Kirkwood Hospice	Money would go towards providing world class specialist care				
10	9	360G-COOPGROUP-11517	Grant to Arthur Rank Hospice Charity	We want to buy specialist equipment, including chair clings an				



Step 2

Categorise grants using the Google Cloud Natural Language API







Step 3

Save labelled results to BigQuery database and export to CSV for analysis



# **Labelling/Categorization of Grants**

Part 2

After categorising 12021 grants, 7943 were given a category, 4078 were unclassified.

Using **Google AutoML**, we can train a supervised classifier on our labelled grants to label the **4078** unclassified grants.

#### 7943 labelled grants

Use labelled grants to train machine learning classifier



#### 12021 labelled grants

Export labelled grants to CSV for analysis

#### Google AutoML Classifier

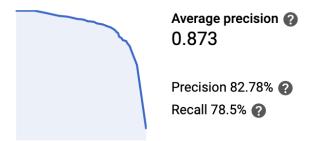
Use trained model on the unclassified grants to obtain labels

## **Trained Classifier for Unlabelled Data**

Part 3



#### grant\_data\_model



Precision and recall are based on a score threshold of 0.5.

**88%** of **Health** grants in the test set were **correctly** classified

**87%** of **Jobs and Education** grants in the test set were classified **correctly** 

However, **53%** of **Travel and Tourism** grants in the test set were incorrectly classified under **People and Society** 

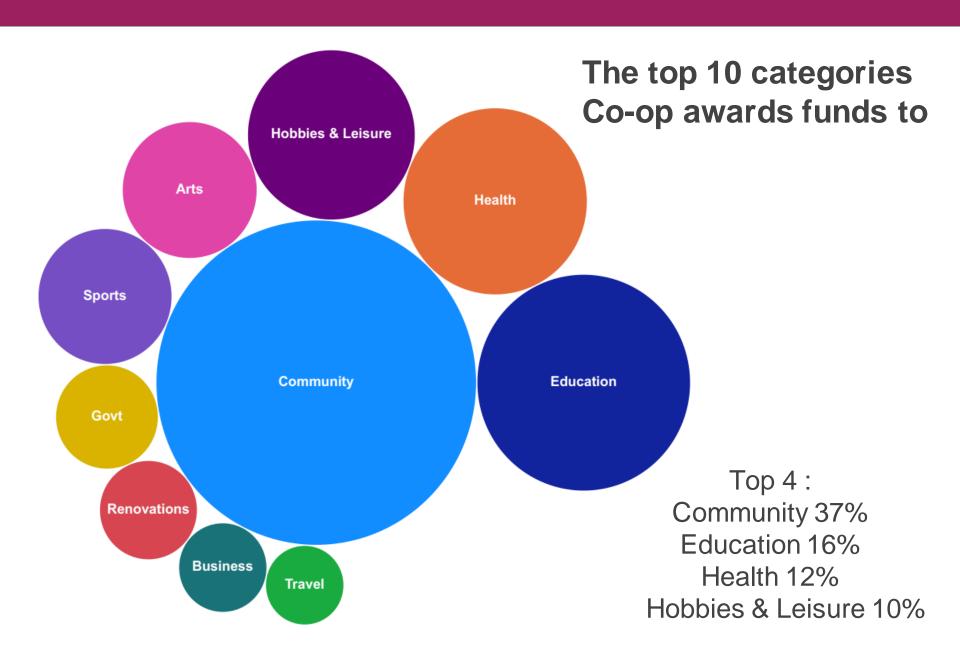
## Precision of 83% Recall of 79%

Overall, classifier returned more correct classifications than incorrect ones on the testing set

#### **Confusion Matrix**

True label	Predicted and the and the teaching and Renoration Hoodness and seems and See								upport Group		
Arts and Entertainment	75%	2%	-	4%	2%	12%	-	-	6%	-	-
Gardening and Renovation		59%		4%	4%	33%					
Health	-	-	88%	1%	1%	10%	-	-	-	-	-
Hobbies and Leisure	-	1%	1%	77%	2%	17%	1%	-	-	-	-
Jobs and Education	2%	-	-	1%	87%	10%	-	-	-	-	-
People and Society	2%	1%	2%	3%	3%	86%	0%	-	2%	-	-
Pets and Animals	-	-	-	57%		43%	-	-	-	-	-
STEM	-	-	-	-	50%	50%	-	-	-	-	-
Sports	-	-	2%	-	-	20%	-	-	78%	-	-
Support Groups	-	-	33%		-	33%		-	-	33%	
Travel and Tourism	12%	-	-	12%		53%	-	-	-	-	24%

# What themes of grants are awarded by The Co-op?



# Where are the local causes located?



# Is there any geographical variation in the amount awarded to recipients based on theme?



#### Recipient.Org.Name

- Alzheimer Scotland- North Aberdeenshire Services
- Arbroath Lifeboat Station
- Barracuda Explorers
- Co-op Foundation
- Monmouth Parkinson's Support Group
- Swanage And Purbeck Development Trust
- The British Red Cross
- Wadebridge Girlguiding
- Waltham Abbey Woodcraft Folk
- West Mersea RNLI

# Next steps...

 Assess labelling and re-train a classifier for our data. Use this new classifier to label remaining data points



 Explore the relationships between label/theme, geographical location and amount awarded



 Create visualisations and draw conclusions of the results based on the research questions

What are the themes of grants that The Co-op awards?

- Where are the local causes located? Is there correlation between themes and locations?
- Is there any geographical variation in the amount awarded to recipients based on theme?

# Thanks!