

An abstract network diagram featuring a complex web of interconnected nodes and lines. The nodes are represented by small circles in various colors including orange, teal, purple, yellow, and grey. The lines are thin and grey, creating a dense, overlapping pattern. The background is a light grey with scattered small black dots.

# PREDICTING INTERSTATE AFFINITY USING MACHINE LEARNING

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AREA/TOPIC

- Inspired by Dr.Vito D'Orazio – predicted violent conflict in Africa using machine learning models
- Wanted to generalize globally
- Predict inter-state affinity with machine learning using political, social, economic. and demographic data

## Event/Affinity Data

- Integrated Crisis Early Warning System
  - ICEWS: [Integrated Crisis Early Warning System from Harvard Dataverse](#) from January 1, 1995, to April 11, 2023
  - Affinity score computed from intensity found in event data



## Coded Event Data

Source.Country	Event.Text	CAMEO.Code	Intensity	Target.Country
Ukraine	Use conventional military force	190	-10.0	Ukraine
Ukraine	Use unconventional violence	180	-9.0	Ukraine
Pakistan	Make statement	10	0.0	Pakistan
United Kingdom	Consult	40	1.0	Pakistan
Pakistan	Consult	40	1.0	United Kingdom
Taiwan	Grant diplomatic recognition	54	6.0	Taiwan
Taiwan	Conduct hunger strike	142	-6.5	NA
Thailand	Arrest, detain, or charge with legal action	173	-5.0	Thailand
China	Praise or endorse	51	3.4	Occupied Palestinian Territory
North Korea	Express intent to meet or negotiate	36	4.0	South Korea

## Socioeconomic Data

- World Bank
  - [World Development Indicators](#)
  - [Education Statistics](#)
  - [Gender Statistics](#)
  - [Health Nutrition and Population Statistics](#)
  - [Doing Business](#) (Objective measures of business regulations and their enforcement)
- International Monetary Fund
- Freedom House
  - Trade data
  - Freedom scores

## METHODS

### Dependent Variable

- Continuous – affinity score from coded event data
- Categorical – affinity scored converted into categories from -1 to 3

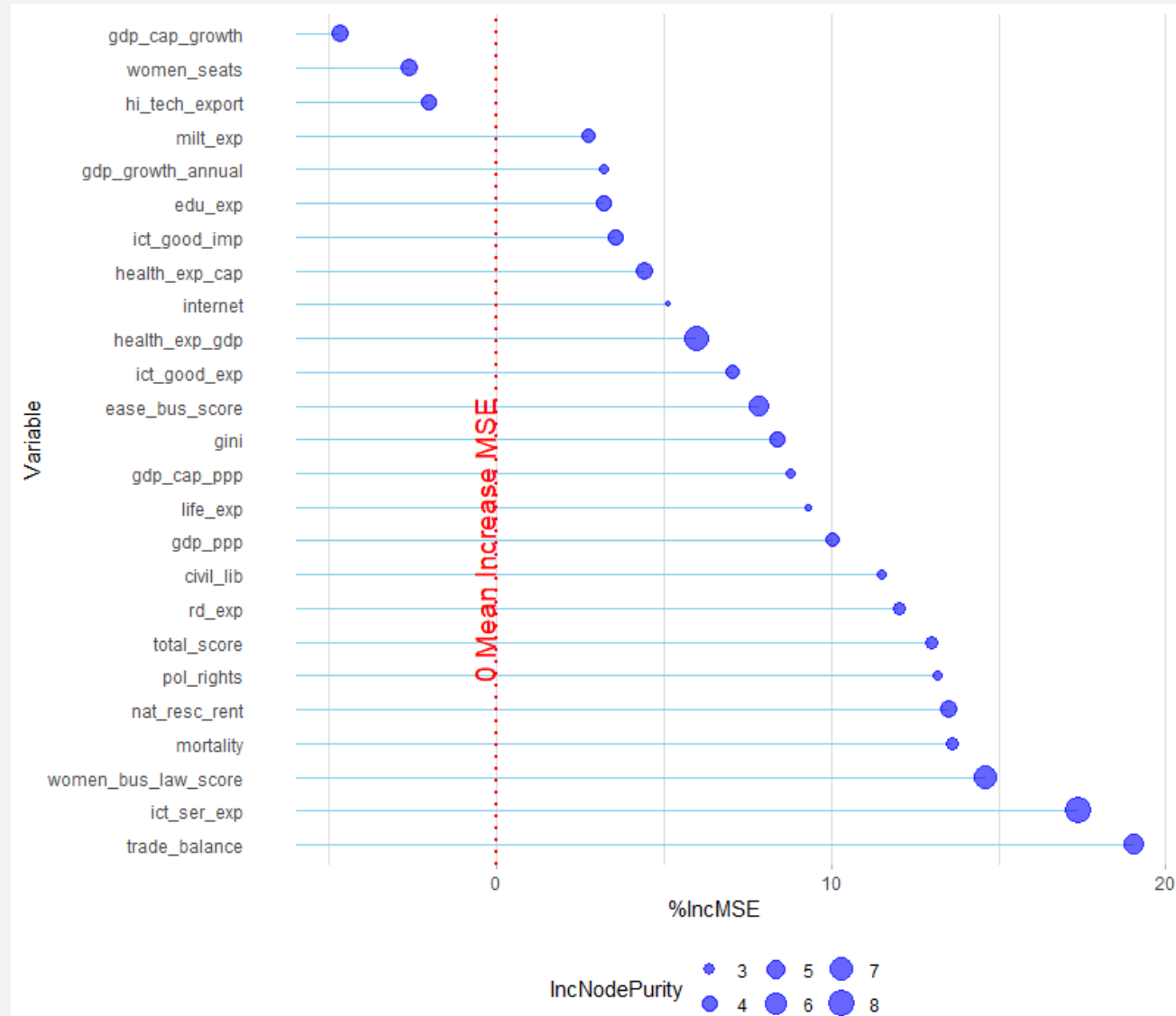
### Random Forests

- Continuous – Regression
- Categorical – Classification

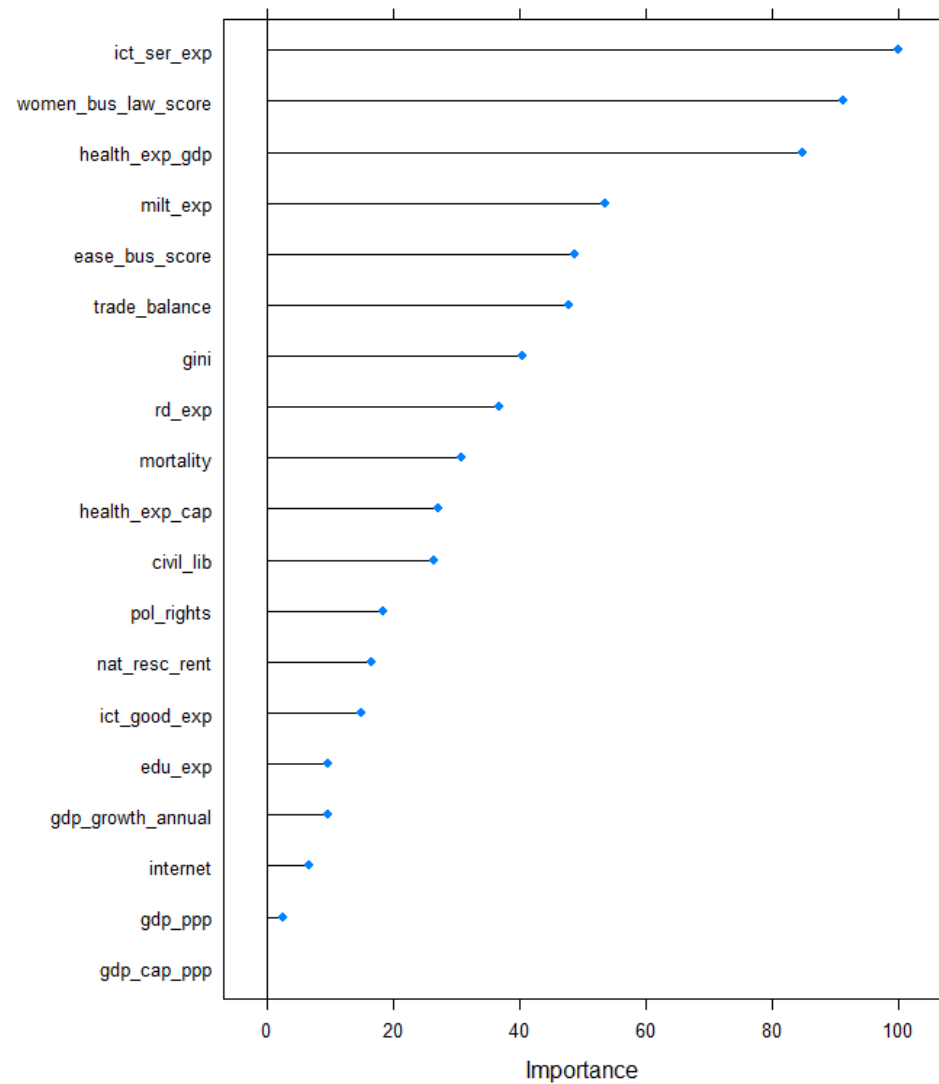
### Deep Learning

- Categorical – TensorFlow classification
- Possible overfitting and little insight using shap library

# Random Forest Regression

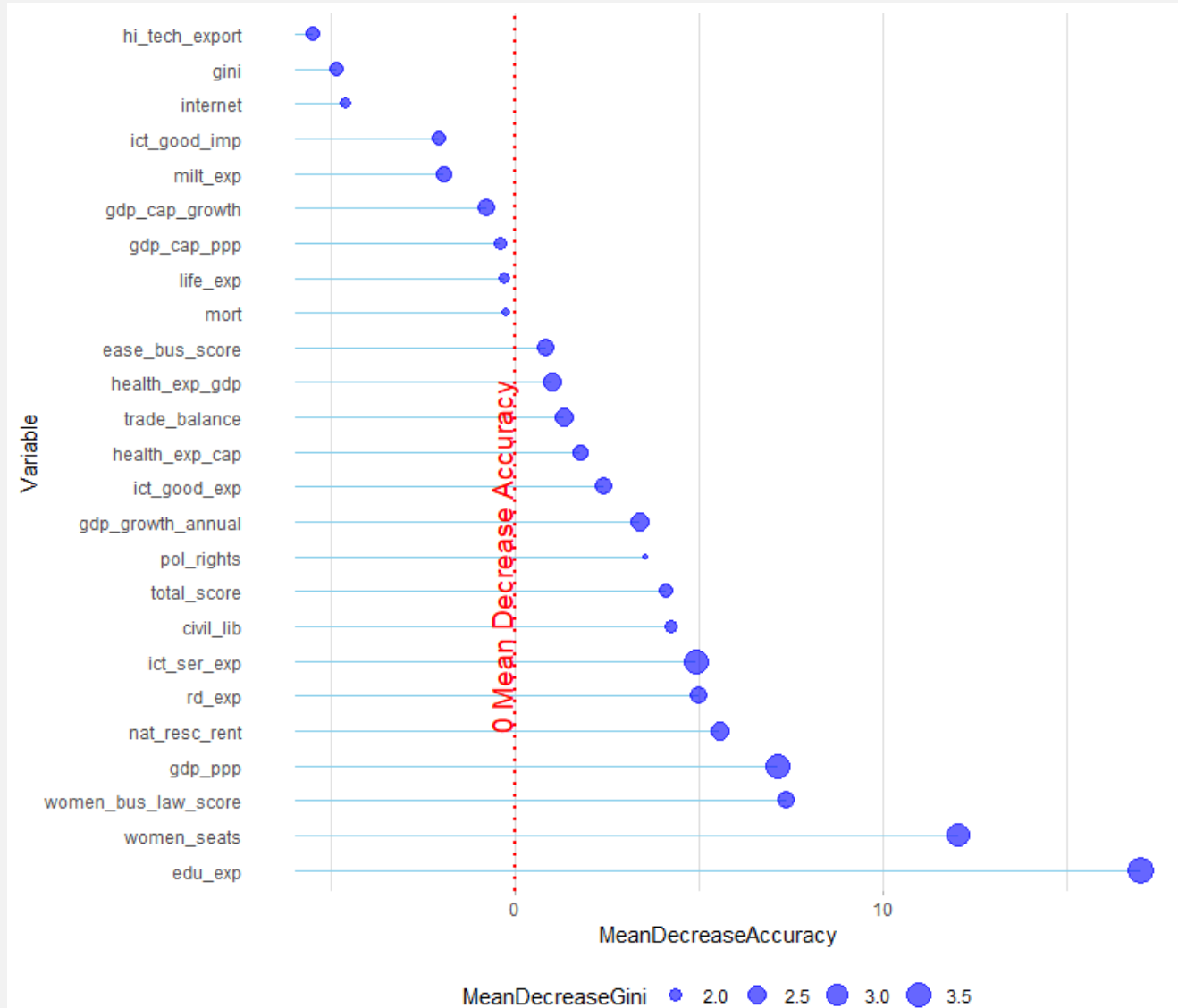


# Random Forest Regression: $R^2 = 0.27$

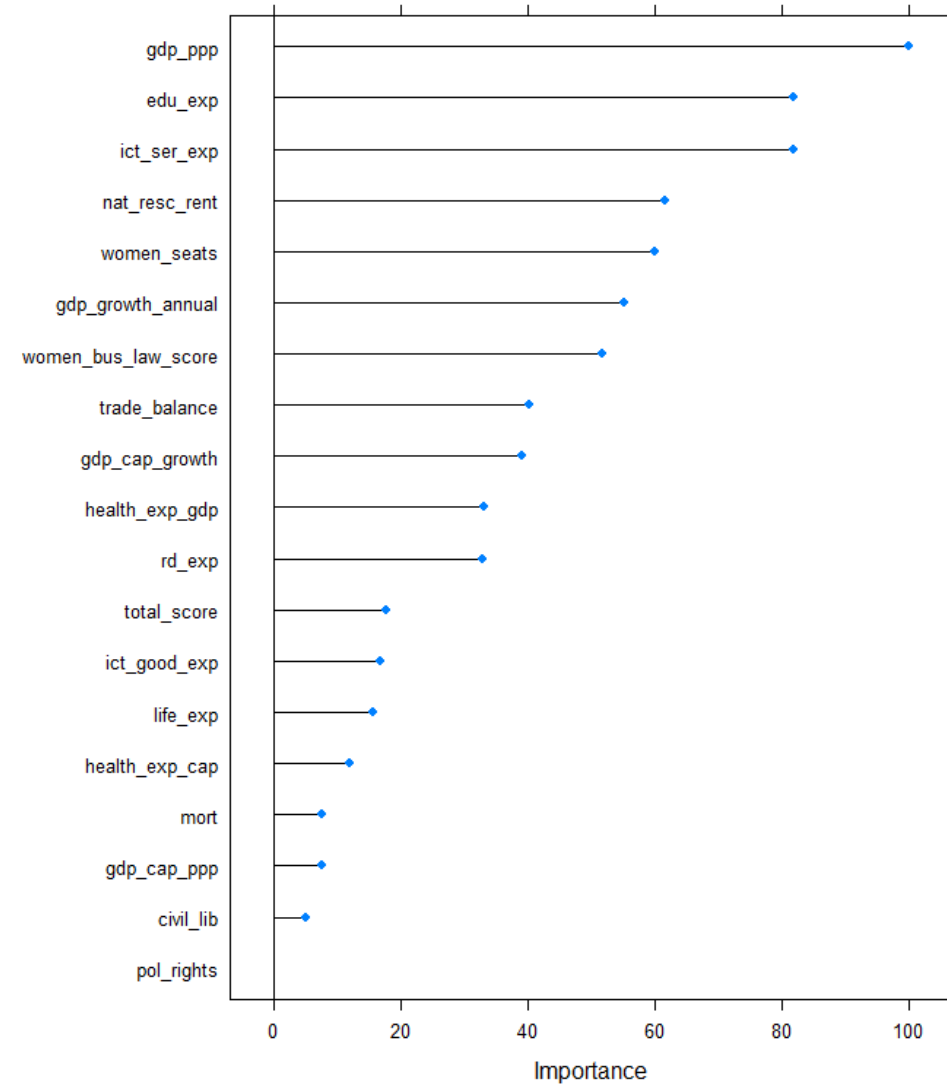




# Random Forest Classification

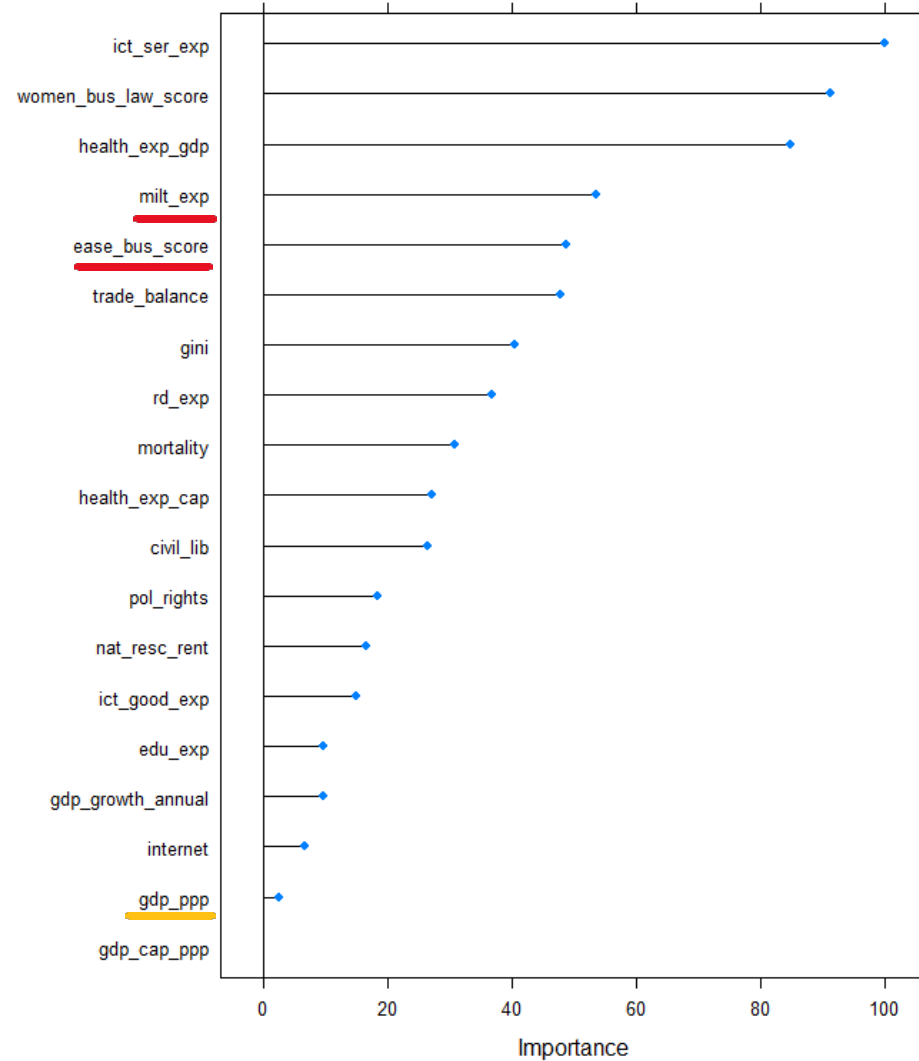


# Random Forest Classification: $R^2 = 0.43$

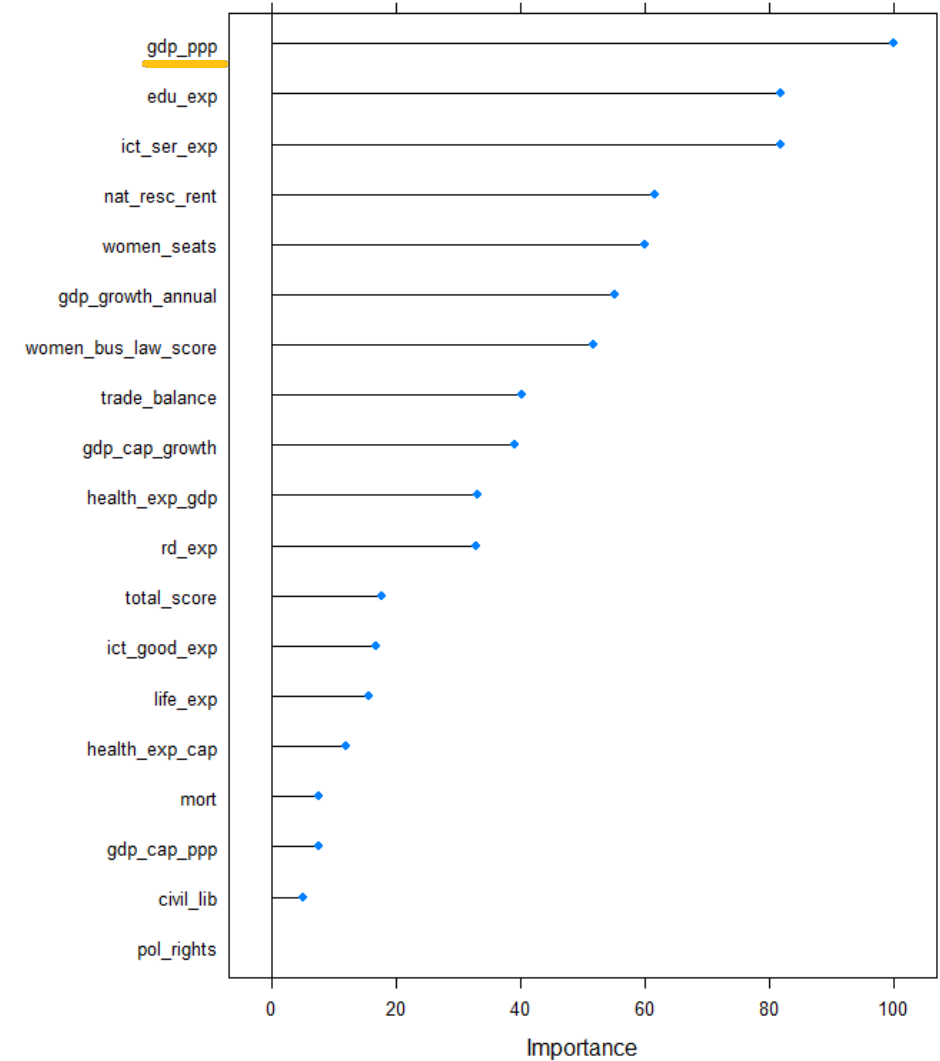


## Variable Importance between Classification & Regression

### Regression



### Classification



## Random Forest Classification: $R^2 = 0.76$

```
Epoch 199/200
11/11 [=====] - 0s 4ms/step - loss: 0.5551 - accuracy: 0.7560
Epoch 200/200
11/11 [=====] - 0s 4ms/step - loss: 0.5550 - accuracy: 0.7560
3/3 [=====] - 0s 5ms/step - loss: 0.6478 - accuracy: 0.7619
Test accuracy: 0.761904776096344
```

```
Received a label value of 4 which is outside the valid range of [0, 3).  Label values: 2 3 3 1
3 3 3 3 3 3 3 3 3 3 3 3 3 3 1 3 4 2 3 3 3 4 4 3 2 3 0
```

## Future Improvements

### Current Model

- Simple affinity calculation
- ICEWS Coded Event Database
- Imputed independent variables and scope of variables



### Improved Model

- Clustering/Network methods to derive affinity communities (Pauls et al., 2017)
- POLECAT Database
- New sources of socioeconomic and demographic data