
Predicting Terror Attacks? A Network Story

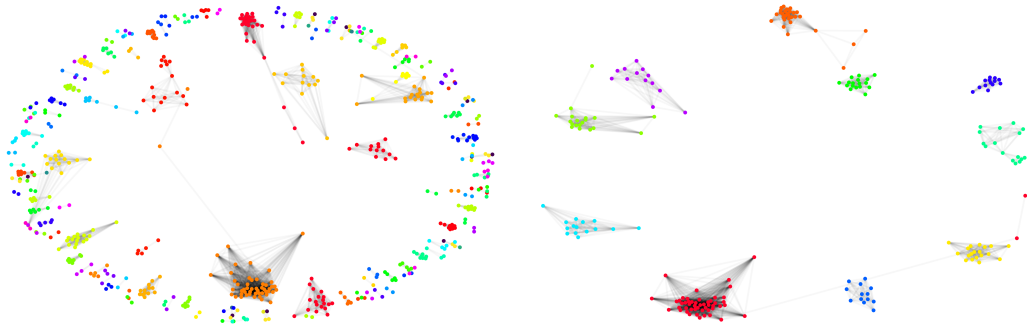
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1 Introduction

2 Exploring the Data

2.1 Relationships Dataset



(a) Terror attacks location graph, colouring by component ID (b) Ten biggest components from the terror attacks location graph

Figure 1: *Graphs analysed in the project*

3 Data Quality

3.1 Terror Attacks Dataset

Multiple issues regarding data quality have been found in this dataset:

Broadness The dataset comprises attacks ranging from 1969 to 1950 and spanning the entire globe. Simple and relevant explanations for the graph formation or properties are not likely to be found, since the mechanisms behind two different attacks can be entirely different.

Structure Half of the nodes are isolated, hence the topological information they carry in the graph is very limited. What is more, because of the transitivity relation described ?????, connected components are in most of the cases complete, hence isotropic.

Reliability Errors have been found in the data. For example nodes `Djibouti_Youth_Movement_19900927` and `Armed_Islamic_Group_19950711` have been connected, whereas the first attack took place in Djibouti [?] and the second one in Paris [?]. Hence algorithms using the data must tolerate some error in order to avoid overfitting.

4 Predictions

Table 1: *Prediction accuracy for different node distance weightings*

Weighting		Best skewness ζ	Accuracy
Gaussian:	$w = e^{-d^2/\zeta} - e^{-1/\zeta}$	0.01	50.5 %
Log-Exponential:	$w = e^{-d} \log \left(\frac{1+\zeta}{d+\zeta} \right)$	0.1	50 %
Linear:	$w = 1 - d$	N.A.	47 %
Square:	$w = \begin{cases} 1 & d < \zeta \\ 0 & \text{otherwise} \end{cases}$	0.1	43 %

5 Conclusion