# Automated Production of Political Indicators

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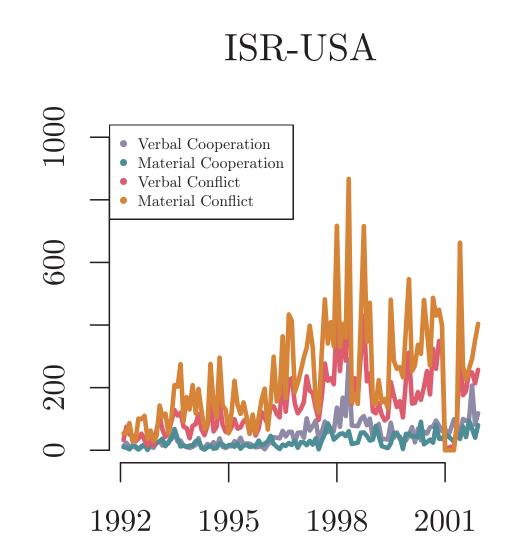


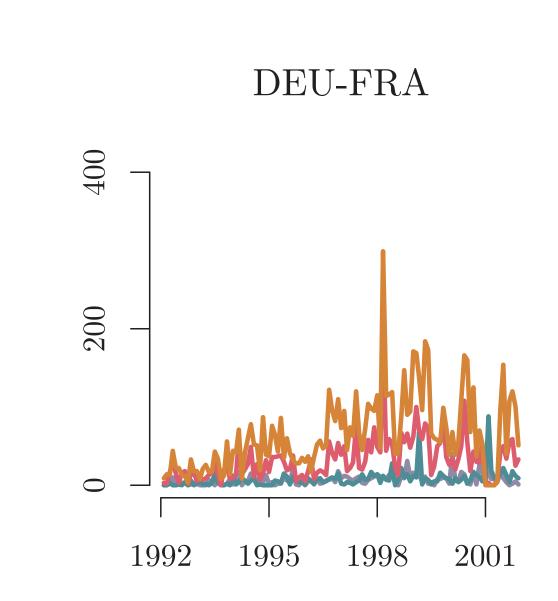
# Motivation

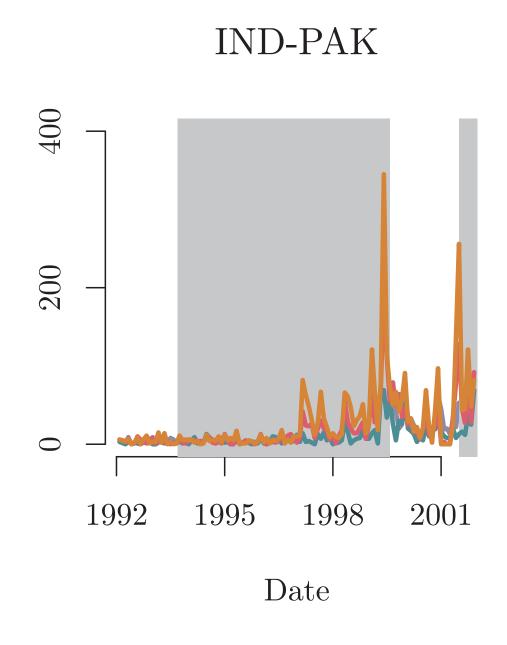
Can political indicators such as the Militarized Interstate Dispute (MID) dataset be approximated by automated classification procedures? Existing, human-intensive pipelines for production of MID and related datasets (e.g. Polity and Freedom House) are costly and slow. This project uses classification trees to estimate the occurrence of MIDs using event data, described below.

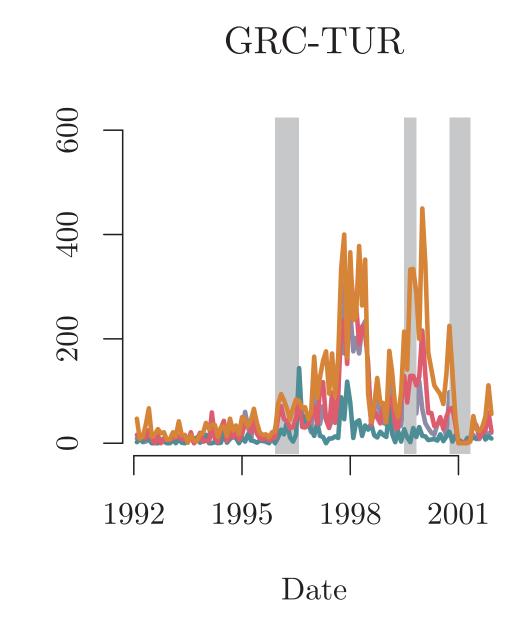
# Data Source

Features were drawn from the Global Database of Events, Language, and Tone (GDELT), which classifies daily interactions as material or verbal, and cooperative or conflictual. Interactions between states were aggregated to the monthly level and first-differenced to account for the exponential growth in the number of records over time. Examples of dyadic time series are shown below, with shaded regions indicating the occurence of a MID.



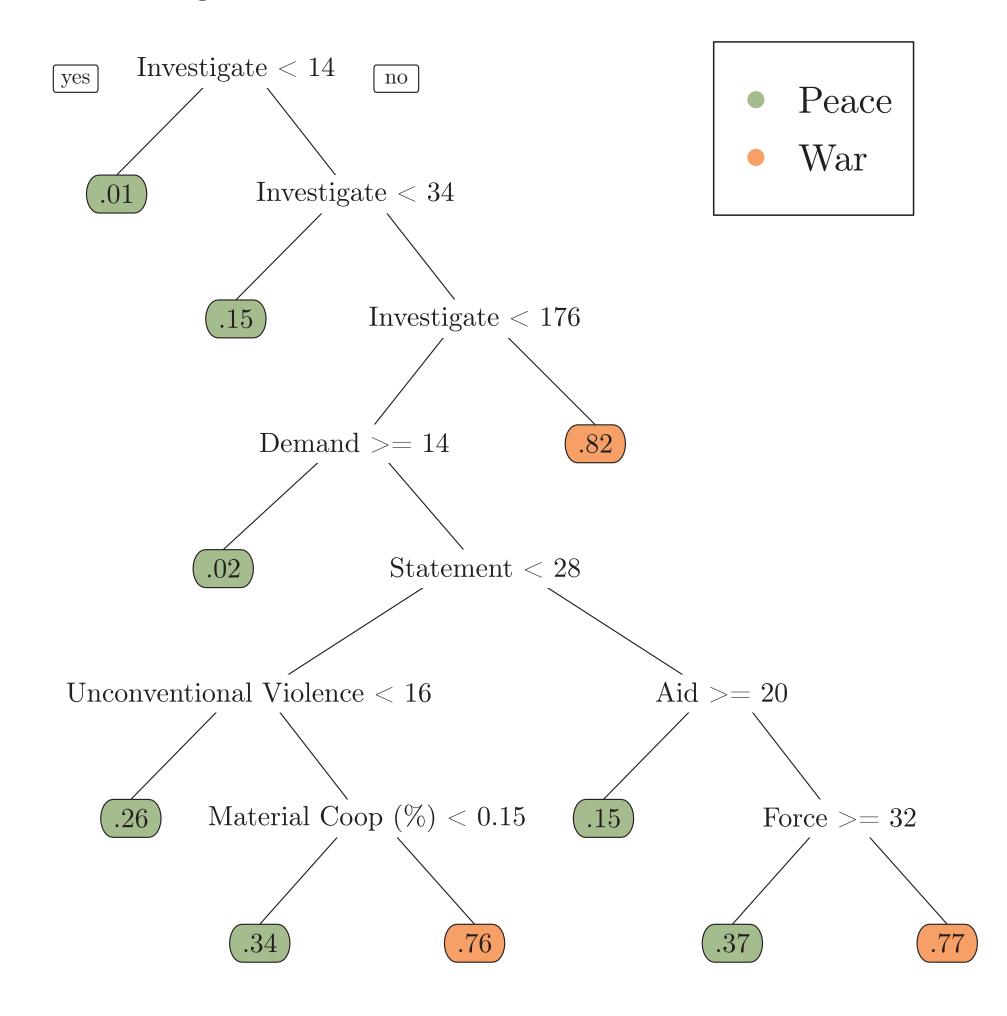






#### Classification Trees

The figure below presents a graphical representation of the final classification tree, fit using the rpart package in R. The minimum complexity parameter,  $\varkappa$ , was set to 0.0001. By ten-fold cross-validation, the optimal  $\varkappa$  was found to be 0.0016, which also minimizes error on the test data. This value was used to prune the tree shown here. The leaves of the tree are shaded according to whether war (MID hostilities  $\geq 4$ ) or peace is more likely, and the value at each leaf indicates the relative frequency of war in the training set.



Each table below presents several measures of accuracy for the classification tree compared to a null model (all predicted values are zero) and a logistic regression model using the same features as the tree. The test set covers 1992 to 1998 (n=372,271) and the training set covers 1999-2001 (n=213,218). The classification tree outperforms the other two models in all respects for the training data, and in most respects for the test data.

Training Data			
Model	MSE	Precision	Recall
Null	0.0075	0.000	0.000
GLM	0.0082	0.158	0.022
CART	0.0067	0.702	0.192
Test Data			
Model	MSE	Precision	Recall
Null	0.0066	0.000	0.000
$\operatorname{GLM}$	0.0079	0.142	0.038
CART	0.0066	0.422	0.027

# Results and Performance

Put in some maps.

# Outlook

- CART handles interactions between features better than GLM.
- Next steps: SMOTE for rare events, use spatial and network lages to account for interdependence between countries, compare to random forest.
- Automated methods can offer a quick and inexpensive first approximation of political indicators.