# Does Polity Score Influence Size of War and Probability of Conflict Escalation?



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# Some work, in two directions

I study interstate wars and conflict, and make attempts at seeing how these are influenced by the Polity IV score. For a given conflict, I use

 $demo = \frac{1}{2} demo_1 + \frac{1}{2} demo_2$ 

as a covariate, in relevant statistical models.

STORY ONE (with Céline Cunen): via Correlates of War (CoW) dataset, we extract

(onset, size, demo) = (x, z, w)

and examine how  $w_i$  influences  $f(z_i | w_i)$ .

STORY TWO (with Jens Kristoffer Haug): via Militarized Interstate Dispute (MID) dataset, we extract lots of pairs of level-of-conflict Markov time series,  $y_1, y_2, \ldots \in \{0, 1, 2\}$ . How do  $w_i$  and other covariates influence  $\Pr(y_{t+1} = 2 | y_t = 1)$ ?

# STORY ONE: Battle deaths, 1823 to 2003

Cunen, Hjort, Nygård (JPR 2020): the World of Wars is not quite stationary; various tests give  $p \doteq 0.04$ ; median left-of-Korea (11,375) bigger than median right-of-Korea (5,240), etc.



# Polity IV scores

We have such scores  $w_i$  for 90 of the 95 wars. The demo score is slowly climbing through time (but also for warring nations):



# Modelling the sizes of wars (with changepoint and covariates)

Céline-Nils invented his three-parameter model, for the distribution of battledeaths above 1000:

$$F(z) = \Pr(Z \le z) = \Big[rac{\{(z-1000)/\mu\}^{ heta}}{1+\{(z-1000)/\mu\}^{ heta}}\Big]^{lpha} \quad ext{for } z \ge 1000.$$

There is power-law behaviour for z becoming large,  $F(z) \doteq 1 - \alpha (\mu/z)^{\theta}$ .

Initial Task: we look for a potential changepoint,  $\tau$ , with  $(\mu_L, \theta_L, \alpha_L)$  to the left of  $\tau$  and  $(\mu_R, \theta_R, \alpha_R)$  to the right of  $\tau$ .

Result: Korea 1950 is the best changepoint candidate: 60 wars to the left have a stochastically larger distribution than the 35 to the right. In our JPR paper we have more, and a confidence curve, etc.

# With Polity IV entering the model

With  $w_i$  the Polity IV (or any other relevant covariate), we let

$$F_i(z) = \Pr(Z_i \leq z_i) = \left[rac{\{(z_i - 1000)/\mu_i\}^{ heta_i}}{1 + \{(z_i - 1000)/\mu_i\}^{ heta_i}}
ight]^lpha \quad ext{for } z_i \geq 1000,$$

with

$$\mu_i = \begin{cases} \mu_{L,0} \exp(\beta_L w_i) & \text{if } i \leq \tau, \\ \mu_{R,0} \exp(\beta_R w_i) & \text{if } i > \tau. \end{cases}$$

Also,  $\theta_i$  is some  $\theta_L$  for  $i \le \tau$  and some  $\theta_R$  for  $i > \tau$ . That is, we attempt to read off how Polity IV influences F via the basic level parameter  $\mu_i$ . Results:

(i) Korea 1950 is again the best changepoint candidate;
(ii) β<sub>L</sub> is slightly negative (ok);
(iii) β<sub>R</sub> is significantly negative (good news).

#### Interpretation & checking via quantiles

For any quantile level q, like q = 0.50 for the median:

$$\phi(q) = F^{-1}(q) = 1000 + \mu \Big(rac{q^{1/lpha}}{1-q^{1/lpha}}\Big)^{1/ heta}$$

So can study quantile as function of  $w \in (-10, 10)$ , Polity IV:

$$F^{-1}(q \mid w) = \begin{cases} 1000 + \mu_{0,L} \exp(\beta_L w) \left(\frac{q^{1/\alpha}}{1-q^{1/\alpha}}\right)^{1/\theta_L} & \text{before 1950} \\ 1000 + \mu_{0,R} \exp(\beta_R w) \left(\frac{q^{1/\alpha}}{1-q^{1/\alpha}}\right)^{1/\theta_R} & \text{after 1950.} \end{cases}$$

There is a slight quantile decrease before Korea 1950, but a clear significant decrease after Korea 1950.

Lots of wars L before 1950, R after 1950, by log battledeaths. After 1950: predicted medians go down with more democracy; before 1950: not so much.



# STORY TWO: Markov chains for levels of conflict 0, 1, 2

Consider two nations in conflict over time,  $Y_1, Y_2, Y_3, \ldots \in \{0, 1, 2\}$  with 0 = relative peace, 1 = conflict, 2 = very serious conflict.



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We model such a series as Markov chains, with  $3 \times 3$  transition probabilities for  $Y_t$  given what has happened up to t - 1:

$$P(t) = \begin{pmatrix} \pi_{00}(t), \pi_{01}(t), \pi_{02}(t) \\ \pi_{10}(t), \pi_{11}(t), \pi_{12}(t) \\ \pi_{20}(t), \pi_{21}(t), \pi_{22}(t) \end{pmatrix}.$$

We use Dynamical Multinomial Regression Models for these, with covariates x(t) at time t. For row 0:

$$egin{aligned} \pi_{00}(t) &= rac{1}{1 + \exp(x(t)^{ ext{t}}eta_{01}) + \exp(x(t)^{ ext{t}}eta_{02})}, \ \pi_{01}(t) &= rac{\exp(x(t)^{ ext{t}}eta_{01})}{1 + \exp(x(t)^{ ext{t}}eta_{01}) + \exp(x(t)^{ ext{t}}eta_{02})}, \ \pi_{02}(t) &= rac{\exp(x(t)^{ ext{t}}eta_{01}) + \exp(x(t)^{ ext{t}}eta_{02})}{1 + \exp(x(t)^{ ext{t}}eta_{01}) + \exp(x(t)^{ ext{t}}eta_{02})}, \end{aligned}$$

and similarly for rows 1, 2. Of brutal importance:

$$\pi_{12}(t) = \Pr(Y_t = 2 \mid Y_{t-1} = 1)$$
  
=  $\frac{\exp(x(t)^{t}\beta_{12})}{1 + \exp(x(t)^{t}\beta_{11}) + \exp(x(t)^{t}\beta_{12})}.$ 

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# Lots o' work

- (i) Need to extract conflict level time series  $Y_1, Y_2, \ldots \in \{0, 1, 2\}$ , for pairs of nations, from MID data.
- (ii) Need relevant covariates  $x(t) = (x_1(t), \dots, x_p(t))$ , with Polity IV one of these (demo-low, demo-high, their average).
- (iii) Then fit the models; log-likelihood can be worked with, but not easy.
- (iv) Then select among candidate models JK Haug constructed a FIC (Focused Information Criterion) in his thesis.
- (v) For the best models, estimate, assess, test, degreee of confidence, interpret, predict.
- (vi) In particular, examine influences of democracy, like Polity IV.

(Brageløfte: Haug + Hjort write a paper after the summer: modelling and selection methodology for regression Markov chains; applied to conflict chains with Polity IV.)

#### How does $Pr(1 \rightarrow 2)$ vary with Polity IV?

Estimated  $\pi_{12}(t \mid w)$ , for Polity IV level  $w \in (-10, 10)$ , other covariates held fixed.



Again, but different basis scenario: Estimated  $\pi_{12}(t \mid w)$ , for Polity IV level  $w \in (-10, 10)$ , other covariates held fixed.



# Remarks

- We're creating a versatile toolbox for analysing single, multiple, many chains of conflict over time. We may examine influences of particular covariates; we have confidence bands around pertinent curves; we may predict under given sets of circumstances.
- A There are variations in the parametric modelling of *F*(z) = Pr(Z ≤ z), with essentially similar results regarding influence of Polity IV: (i) changepoint (more or less) 1950; (ii) democracy not significant before 1950; (iii) democracy helps, in the sense of smaller wars, after 1950.
- The conflict chains Y<sub>1</sub>, Y<sub>2</sub>,... have lots of zeroes (luckily). Might construct different types of models to reflect this.

- Lots o' work with all the regression models for Markov chains.
- Should run the best models with different democracy indexes

   different components tell different stories.
- Interaction between Markov chains: more modelling.
- ♦ Most crucial transition is  $\pi_{12}(t) = \Pr(Y_t = 2 | Y_{t-1} = 1)$ , but also descalation  $2 \rightarrow 1$  is crucial perhaps with different covariates being more important.

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