

Predictors of AfD party success in the 2017 elections

A Bayesian modeling approach

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Menace to society

Right-wing populism then and now

Causes of 20th century world wars

- Perceived in-group superiority (nationalism, racism, antisemitism)
- (Perceived) injustice and inequality
- Autocrats as political leaders

Source: Kershaw, I. (2016). To hell and back: Europe 1914-1949. New York City, NY: Penguin.

Right-wing populism varies greatly, but...



Source: <https://pixabay.com/photos/audience-crowd-people-persons-828584/> Pixabay
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AfD as a nucleus of the German right-wing movement?



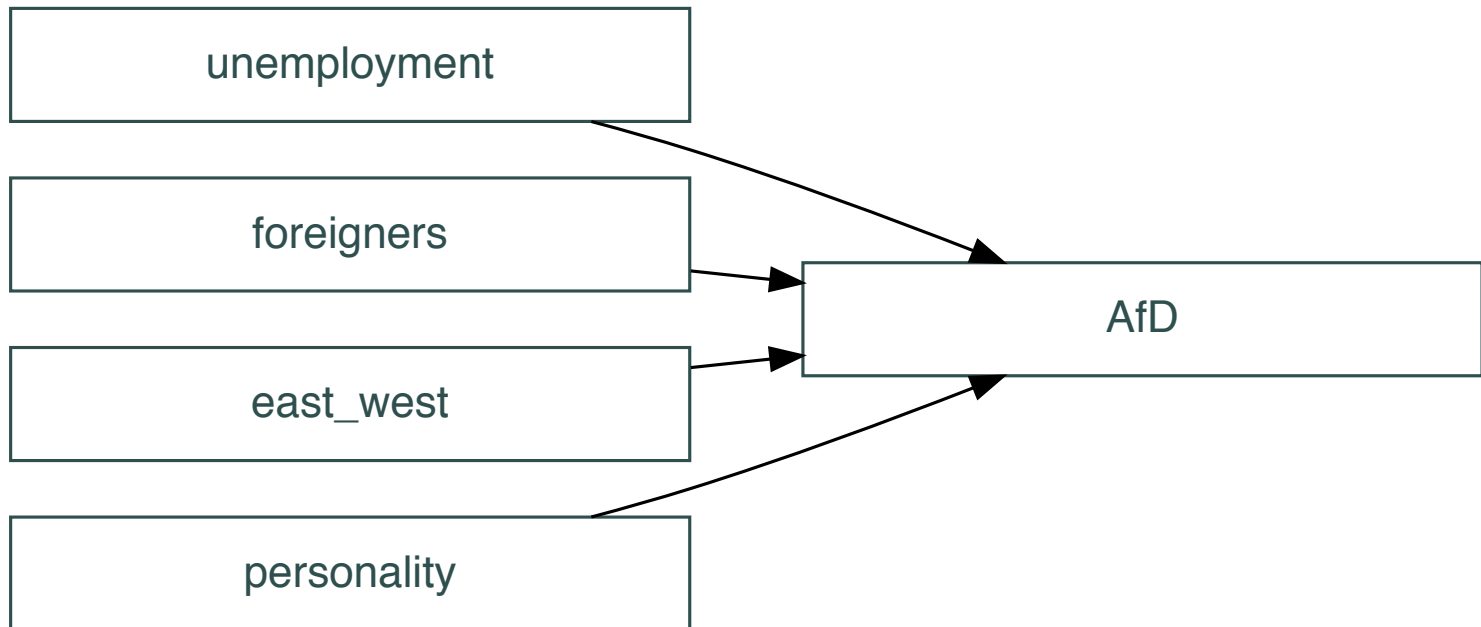
Popular theories on AfD success

Populist party support is fueled by ...

- weak economy ("rust belt hypothesis") 🏭
- high immigration ("flooding hypothesis") 🚶
- cultural patterns ("Saxonia hypothesis") 🚩

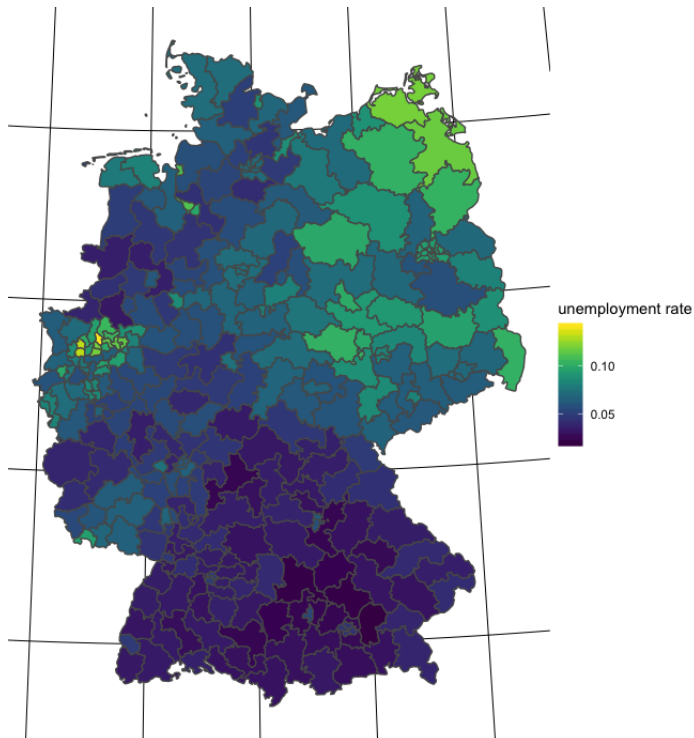
Source: Franz, Christian; Fratzscher, Marcel; Kritikos, Alexander S. (2018) : German right-wing party AfD finds more support in rural areas with aging populations, DIW Weekly Report, ISSN 2568-7697, Deutsches Institut für Wirtschaftsforschung (DIW), Berlin, Vol. 8, Iss. 7/8, pp. 69-79

Our research model

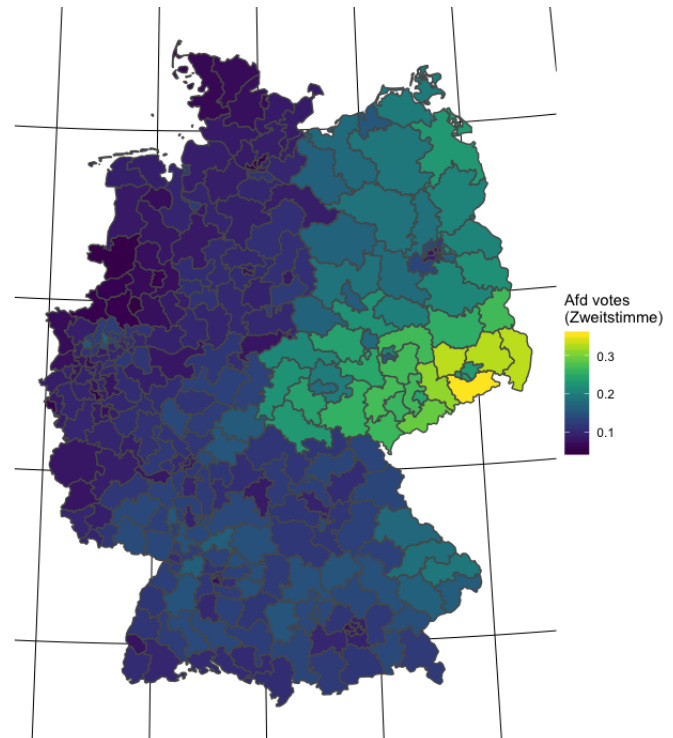


AfD votes, and socioeconomic factors at the Bundestagswahl 2017

Unemployment and AfD votes

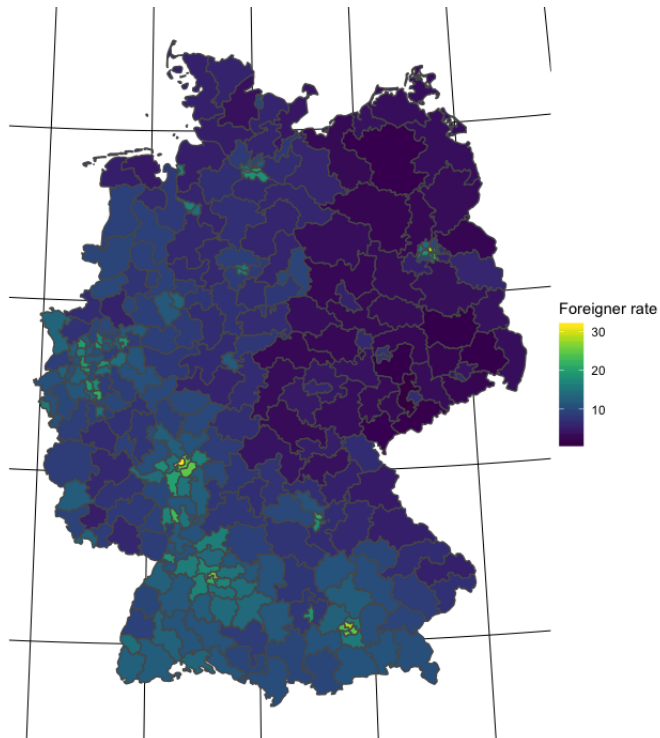


Data provided by the Bundeswahlleiter 2017

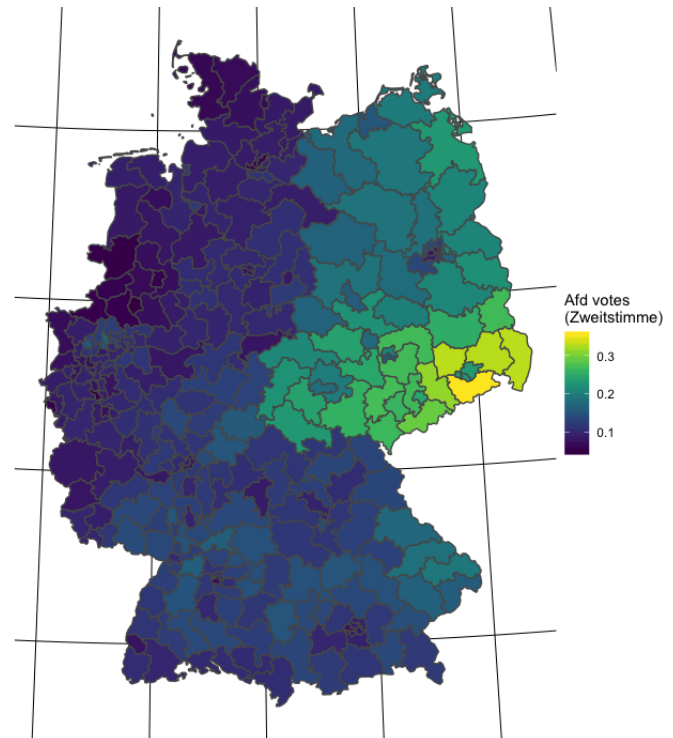


Data provided by the Bundeswahlleiter 2017

Foreigners and AfD votes



Data provided by the Bundeswahlleiter 2017

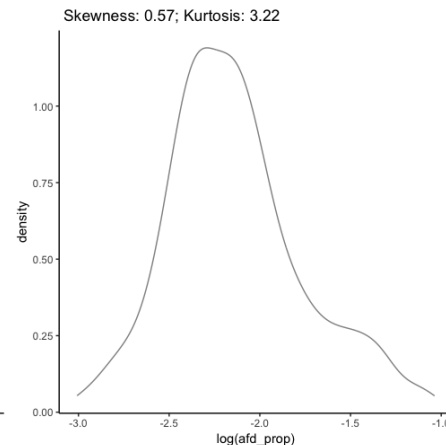
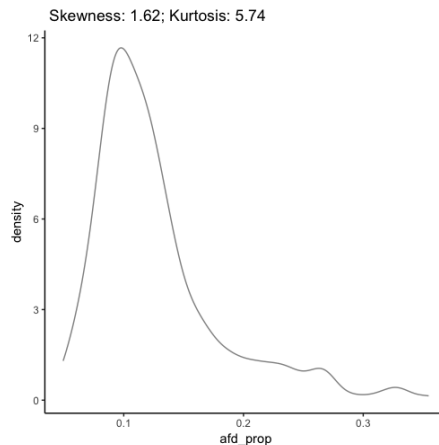


Data provided by the Bundeswahlleiter 2017

data analysis

Data preparation

- Election related data were obtained from [Bundeswahlleiter 2017](#)
- Personality data (n = 12444) were collected by the authors
- Some ZIP codes could not be matched to electoral districts
- Data and analysis are accessible at Github:
https://github.com/sebastiansauer/afd_values
- Outcome variable: `afd_votes` (proportion) was log-transformed for better approximation to normality



Personality model CHOUGHS

- Seven personality types according to CHOUGHS model
 - C onformism
 - H endonism
 - O ut of responsibility
 - U nderstand
 - G ourmets
 - H armony
 - S elf-determined
- Sample: approx. 100k face-to-face interviews (stratified by sex and age)
- Multi dimensional scaling was used to devise types
- CHOUGHS builds on Schwartz' values model

Source: Gansser, O., & Lübke, K. (2018). *The development of new typologies of behaviour based on universal human values and purchasing behavior*, in: Archives of Data Science, Series B, in submission. Gebauer, H., Haldimann, M., & Saul, C.J. (2017). Service innovations breaking institutionalized rules of health care. *Journal of Service Management*, 28(5), 972-935.

Bayes modeling

- Stan via the R package `rethinking`
- Hamiltonian Markov Chain Monte Carlo (MCMC)
- 2000 iterations, 2 chains, 1/2 burn-in
- Multi level regression modeling (varying intercepts)
- The WAIC was used for to compare model performance:
 - is an estimate for *out-of-sample* model performance
 - based on information theory
 - WAIC is similar to the AIC but less restrictive

Guideline: McElreath, R. (2016). Statistical rethinking. New York City, NY: Apple Academic Press Inc.

Model specification

```
# Likelihood:
```

```
afd_prop_log ~ dnorm(mu, sigma),
```

```
# regression:
```

```
mu <- beta0[state_id] + beta1*for_prop_z      + beta2*unemp_prop_z +  
  beta3*enjoyer          + beta4*harmony_seeker + beta5*self_determined  
  beta6*appreciater      + beta7*conformist      + beta8*type_unknown +  
  beta9*responsibility_denier,
```

```
# priors:
```

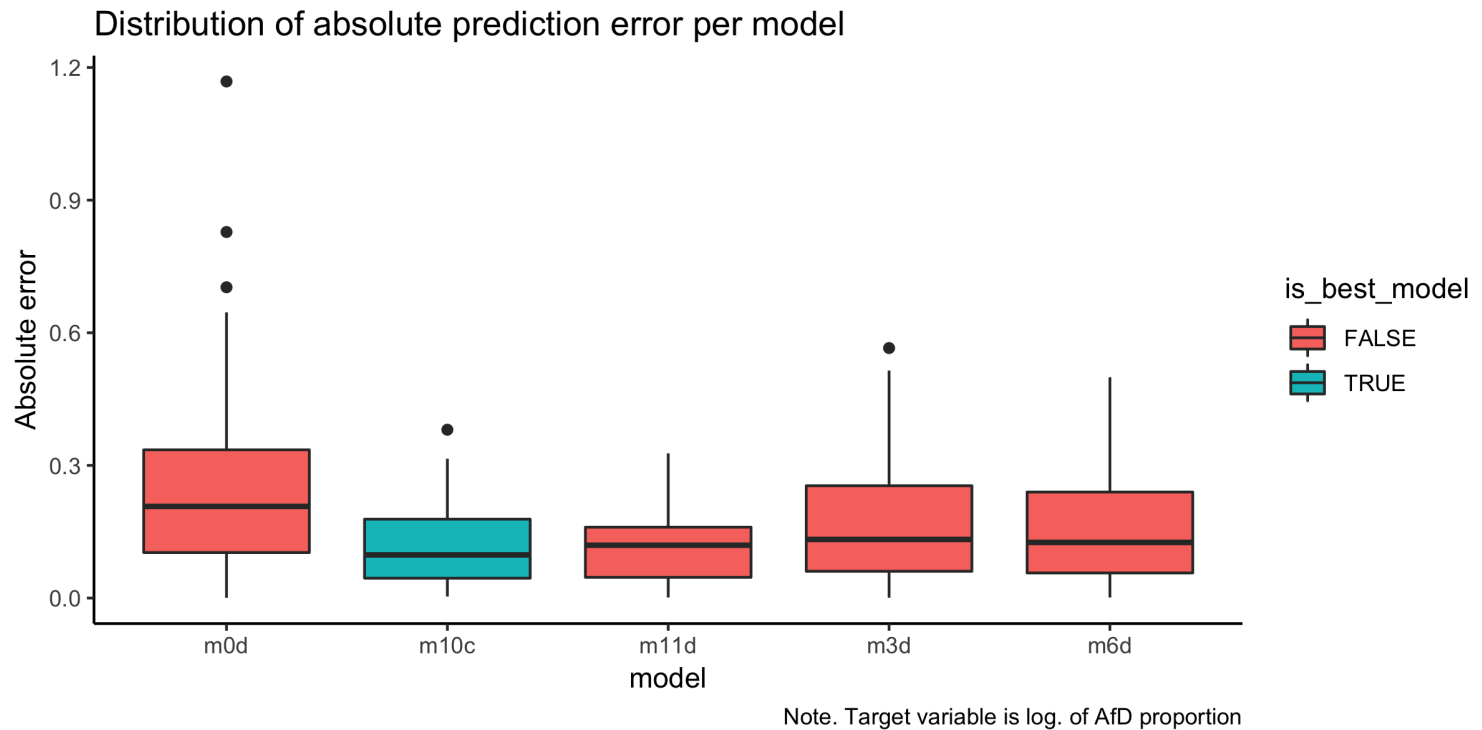
```
sigma ~ dcauchy(0, 1),  
beta1 ~ dnorm(0, 1), beta2 ~ dnorm(0, 1), beta3 ~ dnorm(0, 1),  
beta4 ~ dnorm(0, 1), beta5 ~ dnorm(0, 1), beta6 ~ dnorm(0, 1),  
beta7 ~ dnorm(0, 1), beta8 ~ dnorm(0, 1), beta9 ~ dnorm(0, 1),  
beta0[state_id] ~ dnorm(0, sigma2), # multi level  
sigma2 ~ dcauchy(0, 1)
```

Results: Model comparison

State is the strongest predictor

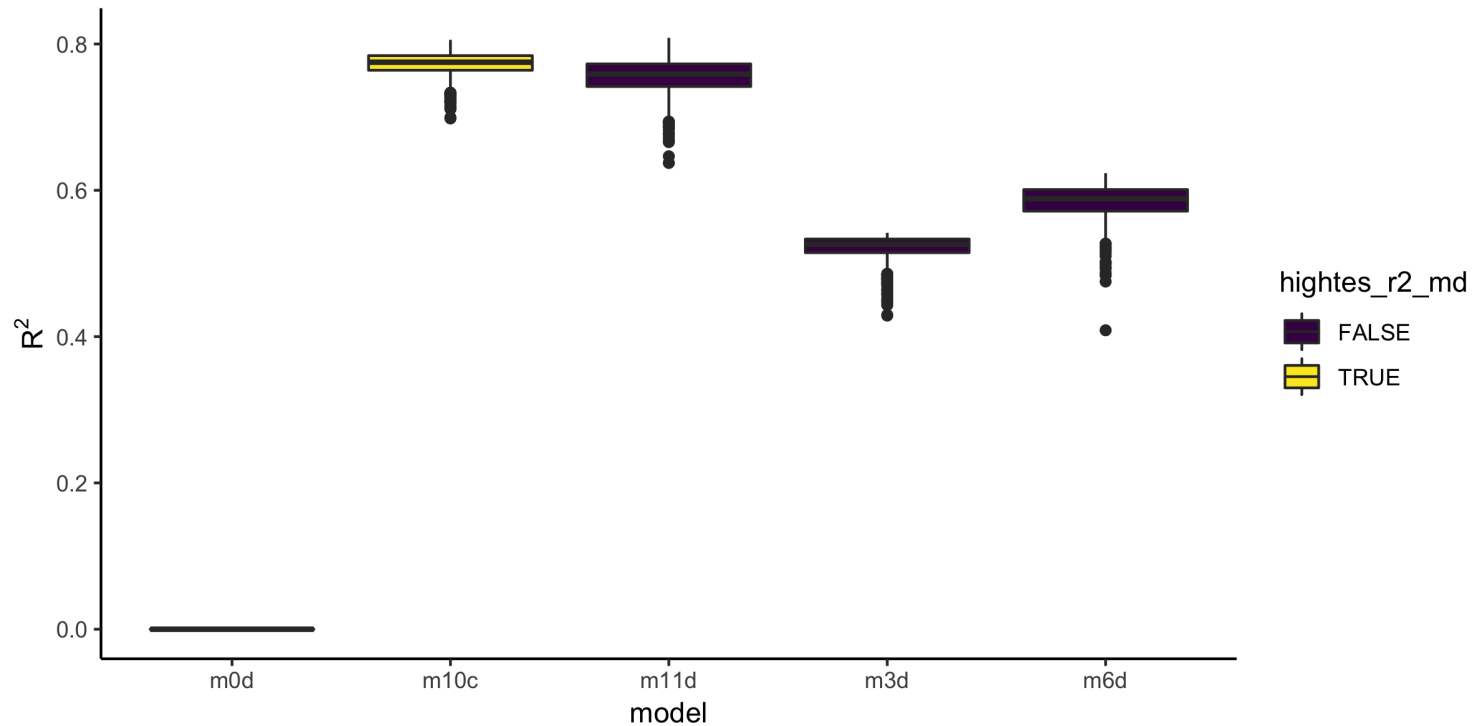
name	predictors	type	WAIC	SE	weight
m10c	unemp, foreign, state	Gaussian	-50.97	10.74	1
m11d	unemp, foreign, state, 8 consumer types	Gaussian	-39.02	10.31	0
m06d	unemp, foreign, east, 8 consumer types	Gaussian	-6.96	12.50	0
m03d	unemp, foreign, east	Gaussian	-1.24	12.44	0
m00d	none	Gaussian	54.39	16.13	0
m12d	unemp, foreign, state, 8 consumer types	Poisson	64311.15	10241.34	0
m09b	unemp, foreign, state	Poisson	64453.60	9016.30	0
m00e	none	Poisson	211670.94	51582.24	0

Comparing model errors



R squared estimates for each model

Beware: Unadjusted R^2 estimates, prone to overfitting



Results: Most favorable model

Model specification of most favorable model

Model predictors: state (as multi level) + foreign + unemp

```
# Likelihood:
afd_prop_log_z ~ dnorm(mu, sigma),

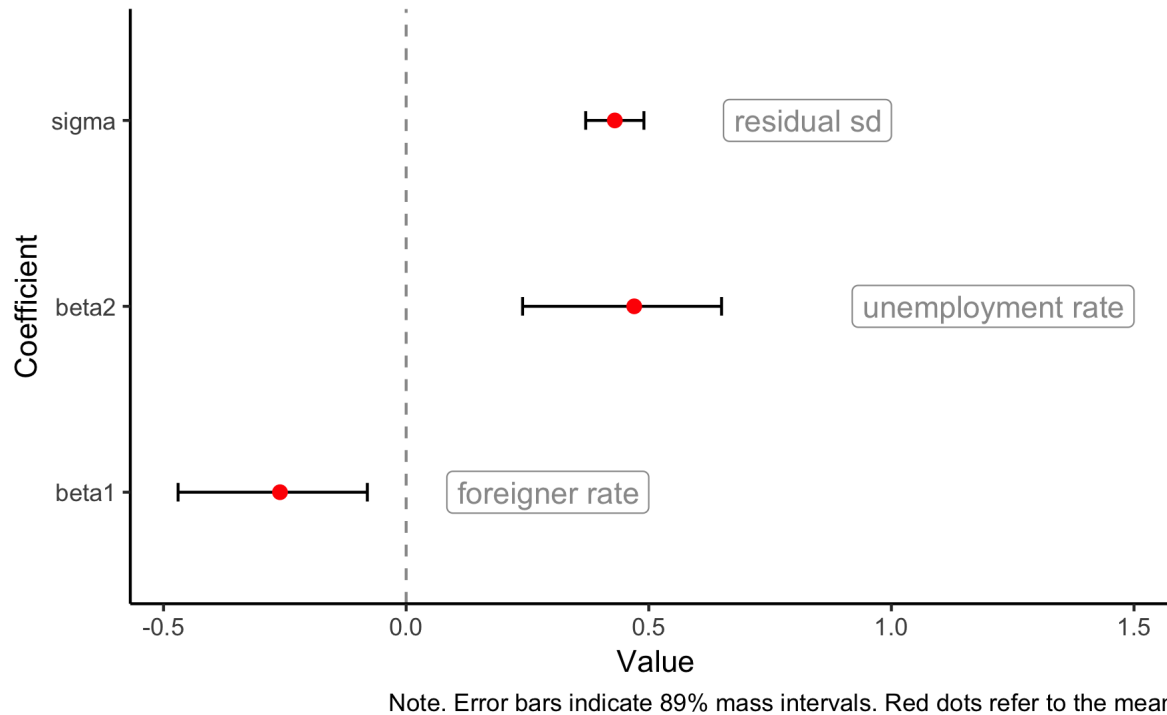
# regression:
mu <- beta0[state_id] + beta1*for_prop_z + beta2*unemp_prop_z,

#priors:
beta0[state_id] ~ dnorm(0, sigma2),

sigma ~ dcauchy(0, 1),
sigma2 ~ dcauchy(0, 1),
beta1 ~ dnorm(0, 1),
beta2 ~ dnorm(0, 1)
```

Coefficients of the most favorable model

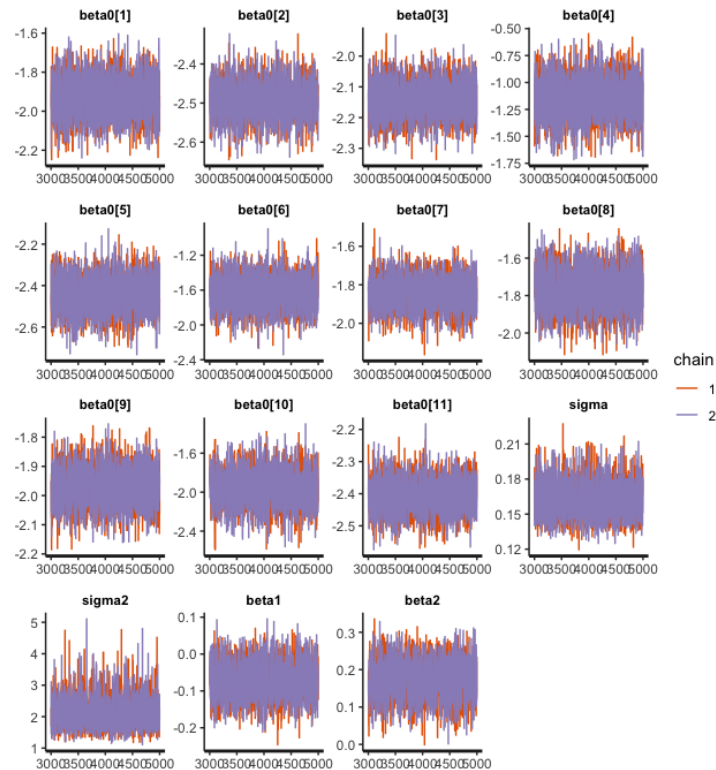
Model predictors: state (as multi level) + foreign + unemp



Only level 1 coefficients are shown.

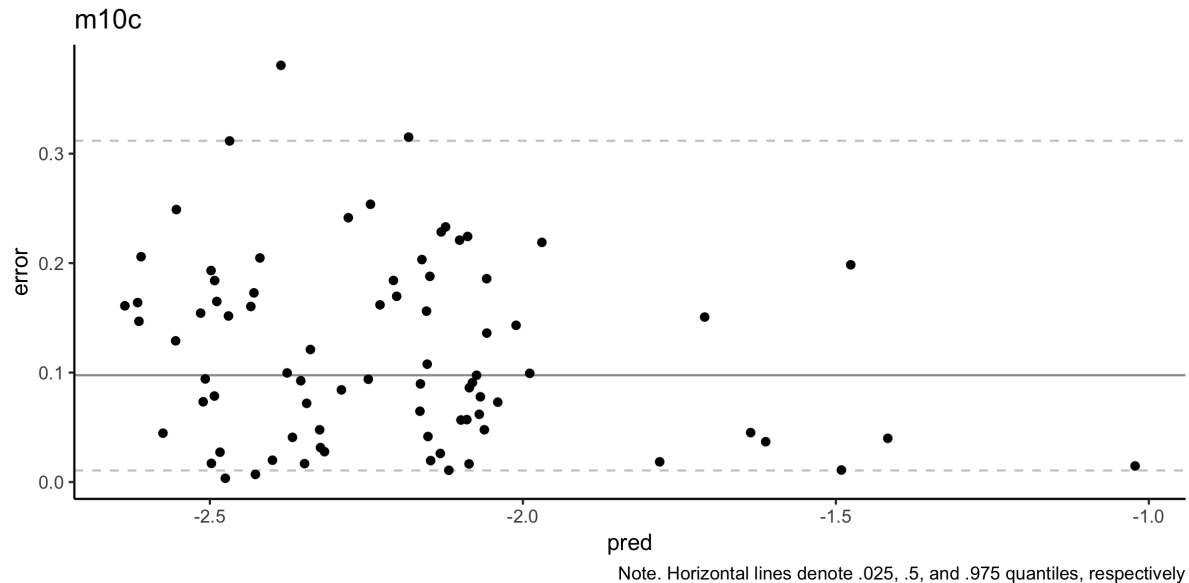
Big fat hairy caterpillars, as it should be

Model predictors: state (as multi level) + foreign + unemp



Model additivity assumption of best model

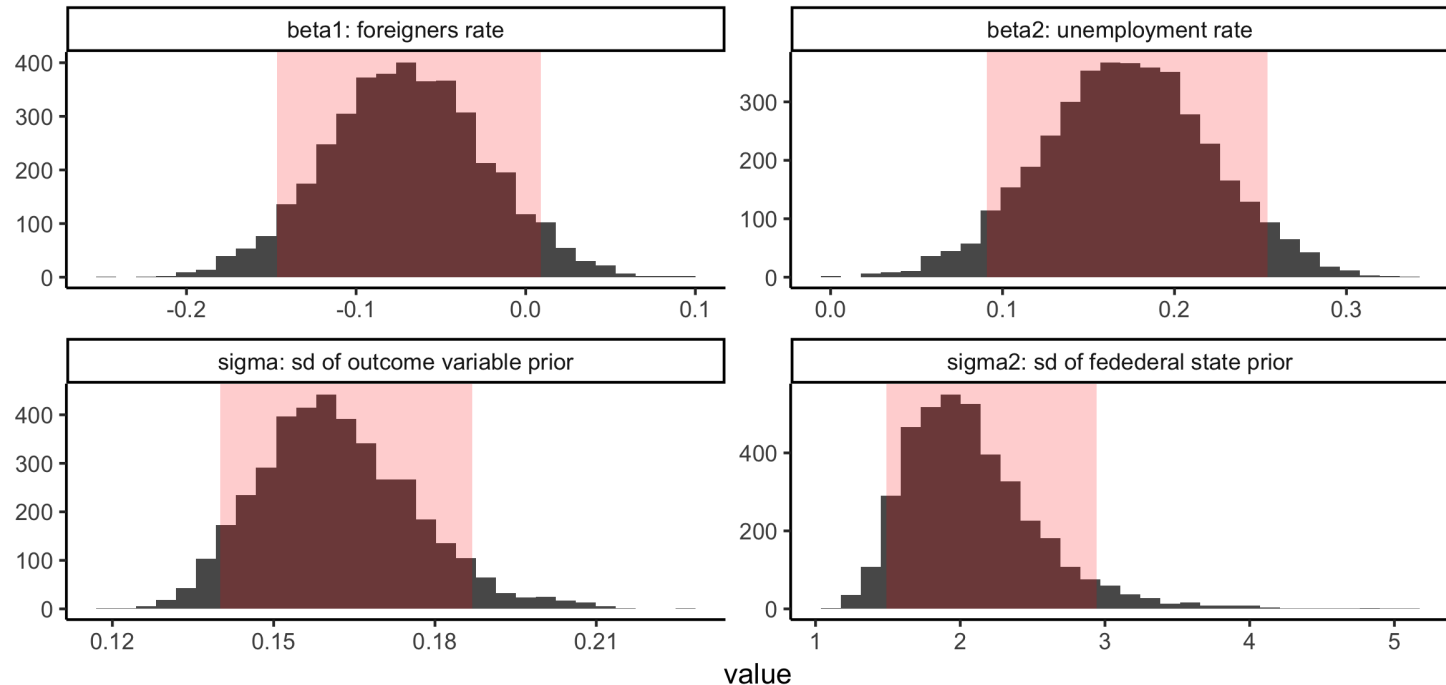
Model predictors: state (as multi level) + foreign + unemp



Gelman, A., & Hill, J. (2006). Data analysis using regression and multilevel/hierarchical models. Cambridge university press.

Posterior distributions of best model

Model predictors: state (as multi level) + foreign + unemp

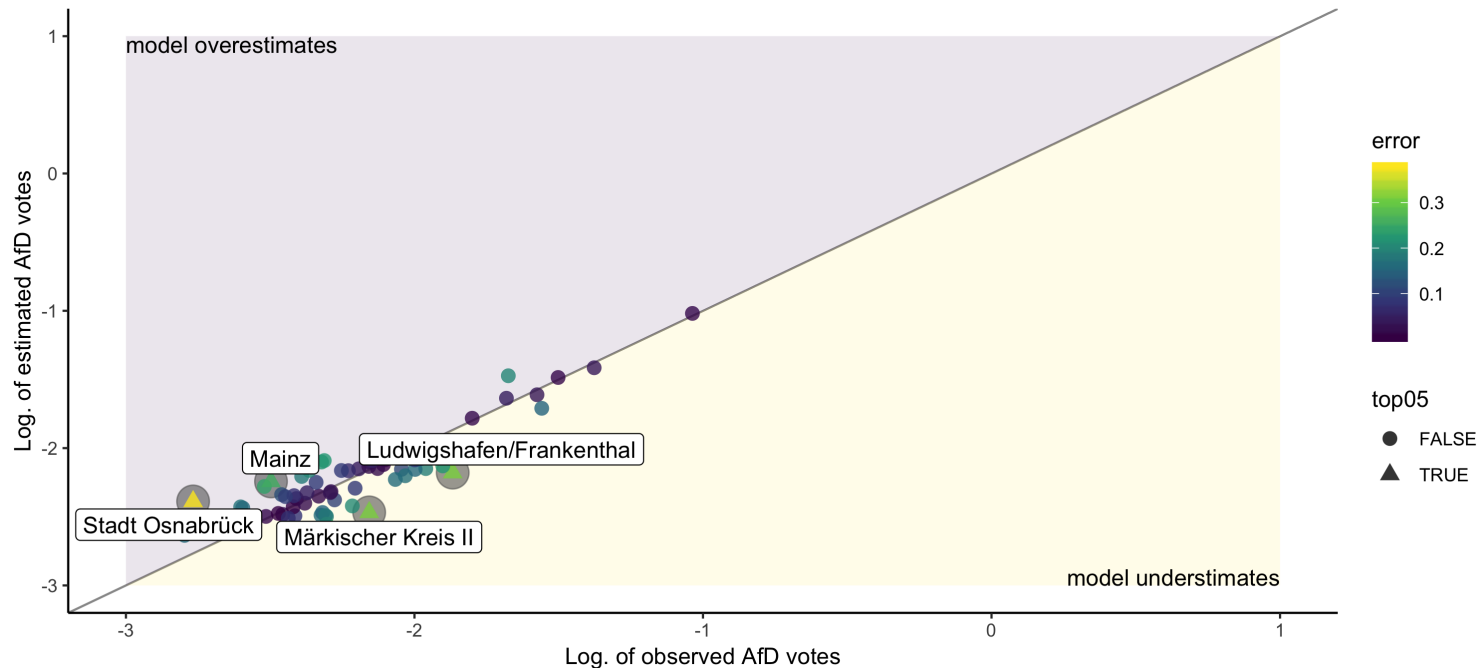


Note. Shaded areas demark 90% mass intervals

Observed vs. estimated AfD votes

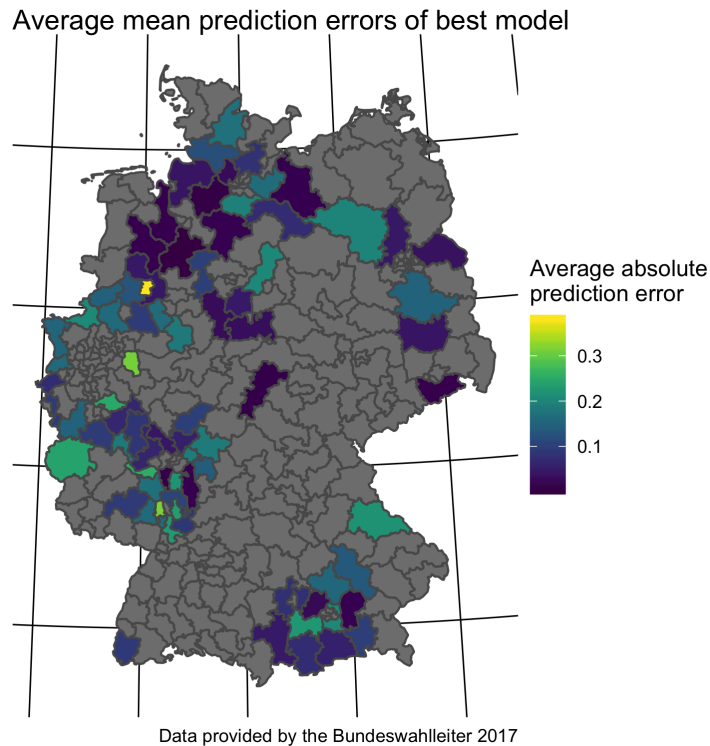
Model predictors: state (as multi level) + foreign + unemp

Modelled vs. observed AfD votes. Top 5 percent predicted errors are labelled



n = 79 electoral districts; data provided by Bundeswahlleiter 2017

Regional patterns of prediction errors (as to best model)



Conclusions

Theoretical implications

- *Region related patterns* appear to play an important role
 - more than unemployment rate and foreigner rate
 - not yet well understood
 - rural? aged society?
- The present model is *simplistic*
- (The proposed) personality pattern didn't show strong impact
- Personality data *representative*?
- Let's model *future elections*
- Pathways of voter behavior remains opaque

Nicole Berbuir, Marcel Lewandowsky & Jasmin Siri (2015) The AfD and its Sympathisers: Finally a Right-Wing Populist Movement in Germany?, *German Politics*, 24:2, 154-178, DOI: 10.1080/09644008.2014.982546


Statistical implications

- *Observational research* is a very *limited* guide for *causal* interpretations
- *Overfitting* (and underfitting) is to be expected
- *Reduced* sample size of electoral districts warrants further investigation
- *Explorative* study, no strong conclusions warranted
- *More models* are possible (but inject researchers' degree of freedom)

Thank you

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 **Get slides here: https://data-se.netlify.com/slides/afd_ecda2019/afd-modeling-ECDA-2019.pdf**

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