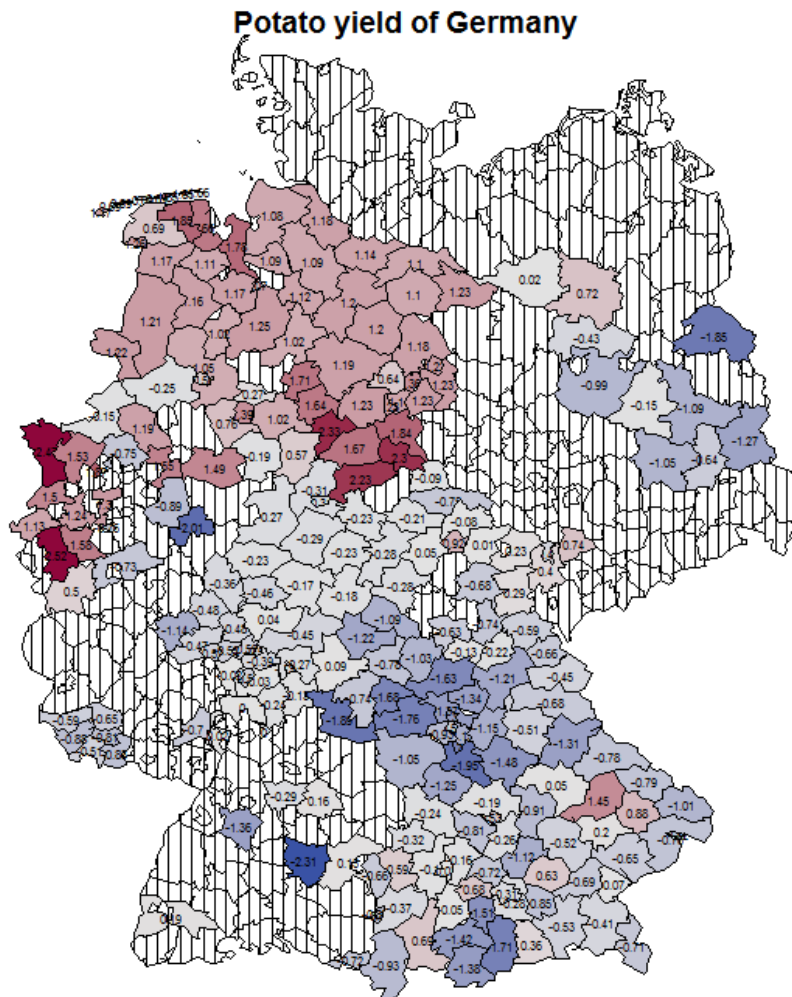




Agricultural yield estimation with spatial smoothing

Caner Erdem

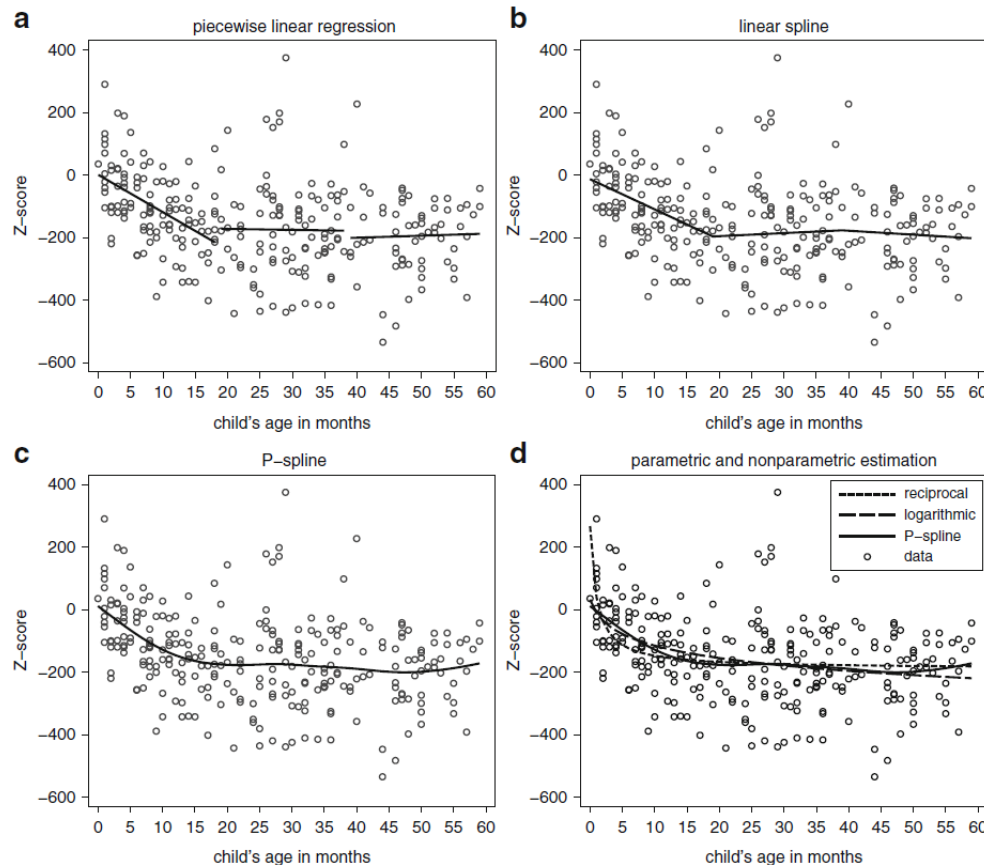
Goal



- Spatial smoothing
- Markov random field
- Penalized least square criterion
- One dependent variable and one covariate

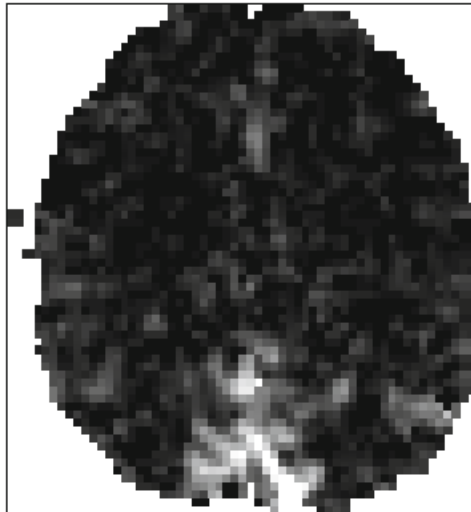
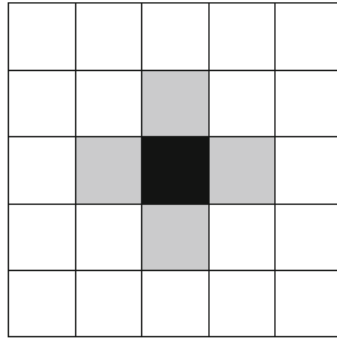
Background - I

- Nonparametric regression
 - Nonlinear effects
 - Difficult to fit parametric models
 - Data-driven & flexible
 - Exploratory data analysis



Source: Fahrmeir et al., 2013

Background - 2

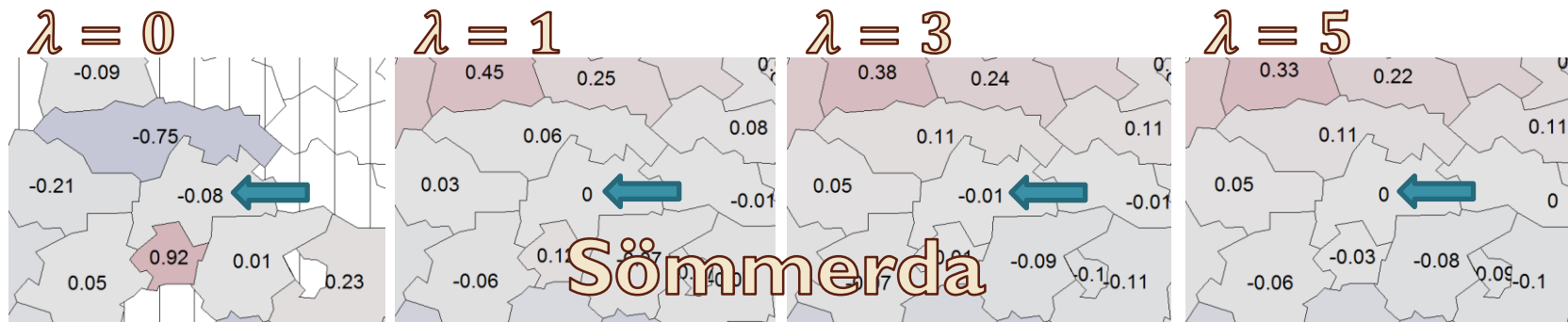


Source: Fahrmeir et al., 2013

- Spatial smoothing
 - Nonparam. method
 - Continuous location variables: geographic coordinates
 - Discrete spatial units: regions or grid
 - Sometimes overlap: human brain mapping
 - Euclidean distance vs. **neighborhood**

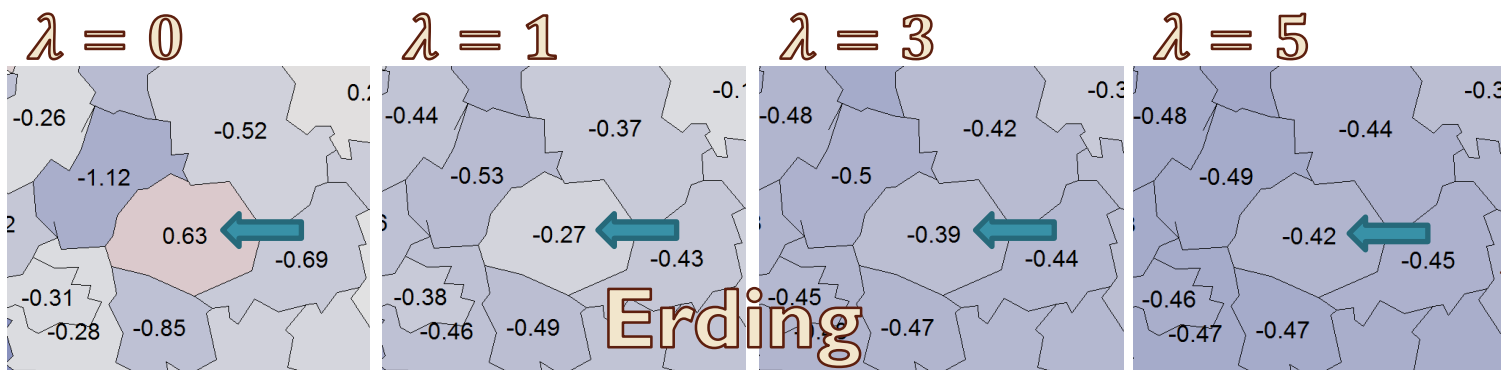
Background - 3

- Markov random field
 - Relationship between regions: neighborhood
 - A graph $G = (V, E)$
 - $V = \{1, 2, \dots, d\}$ associated with γ_s for $s = 1, \dots, d$
 - Neighborhood of s is N_s , $r \in N_s \Leftrightarrow (s, r) \in E$
 - The property: $p(\gamma_s \mid \{\gamma_r\}_{r \in V \setminus s}) = p(\gamma_s \mid \{\gamma_r\}_{r \in N_s})$
 - Information transfer through local connections



Background - 4

- Penalized least square (PLS)
 - $s \sim r$ denotes region s and r are neighbors
 - Regression coeff. $f_{geo}(s) = \gamma_s, s = 1, 2, \dots, d$
 - PLS criterion:
 - $\sum_{i=1}^n (y_i - f_{geo}(s_i))^2 + \lambda \sum_{s=2}^d \sum_{r \in N(s), r < s} (\gamma_r - \gamma_s)^2$
 - Penalty term discourages large deviations in regression coefficients of nearby regions



Background - 5

> GermanyGraph[5:10,5:10]

	1051	1053	1054	1055	1056	1057
1051	4	0	-1	0	0	0
1053	0	7	0	0	0	0
1054	-1	0	2	0	0	0
1055	0	0	0	4	0	-1
1056	0	0	0	0	4	0
1057	0	0	0	-1	0	5

- Model and solution

- Model: $y_i = f_{geo}(s_i) + \varepsilon_i$, $\varepsilon \sim N(0, \sigma^2)$
- Model in matrix notation: $y = Z\gamma + \varepsilon$
- Design mat. $Z[i, s] = \begin{cases} 1 & \text{if } y_i \text{ is in region } s \\ 0 & \text{otherwise.} \end{cases}$
- Penalty mat. $K[s, r] = \begin{cases} -1 & s \neq r, \quad s \sim r, \\ 0 & s \neq r, \quad s \not\sim r, \\ |N(s)| & s = r. \end{cases}$
- PLS estimate: $\hat{\gamma} = (Z'Z + \lambda K)^{-1}Z'y$

Background - 6

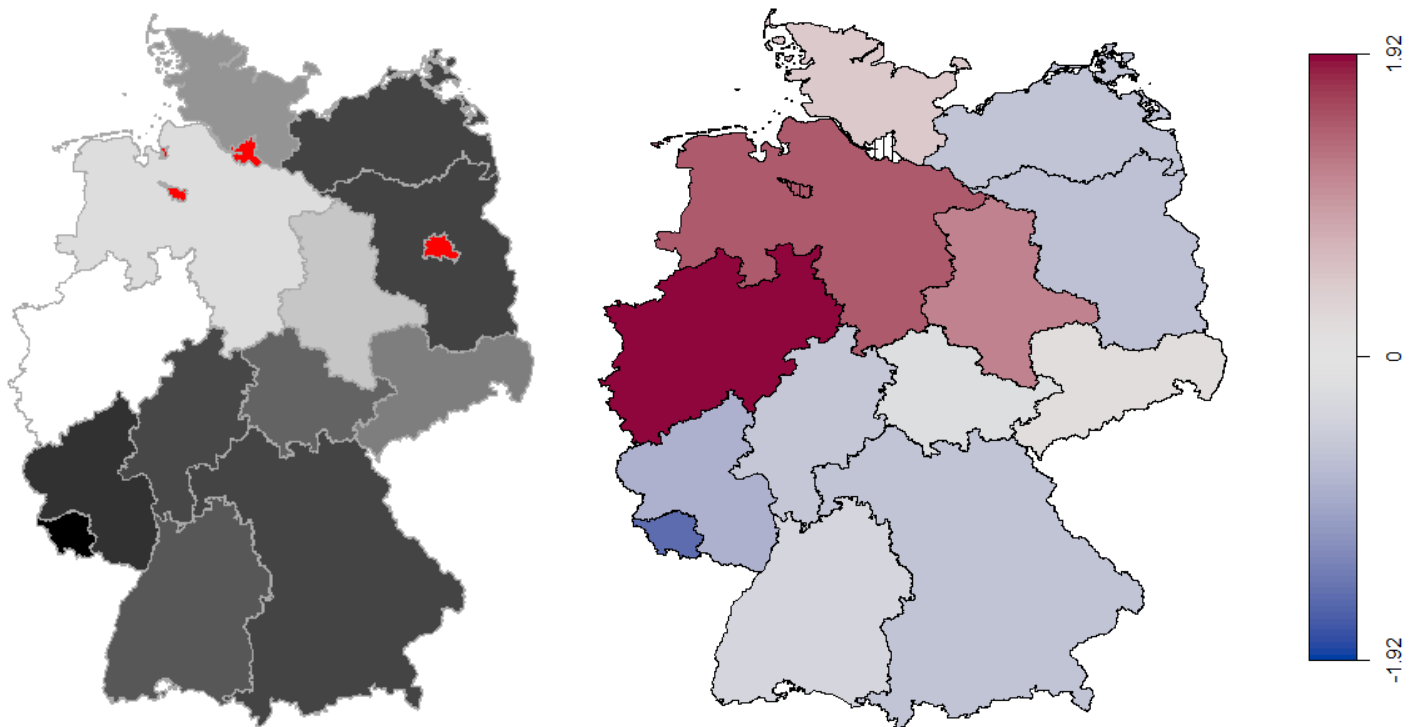
- Concerns about this method
 - Simplicity
 - PLS estimate $\hat{\gamma}$: all in one
 - i.e. spatial and non-spatial covariates
 - Simultaneous est.
 - Mixed models
 - Bayesian approaches
 - BayesX software?



Source: eBay, 2017

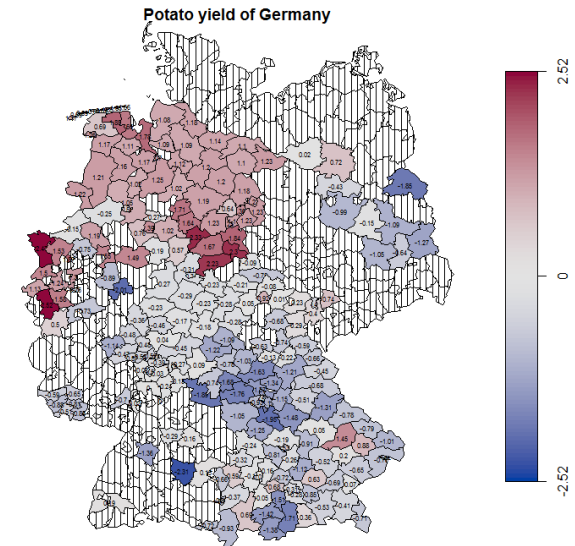
First steps

- state based potato yield (100kg/ha)
- only three states without official data
- gadm.org & regionalstatistik.de



Estimation of spatial effects

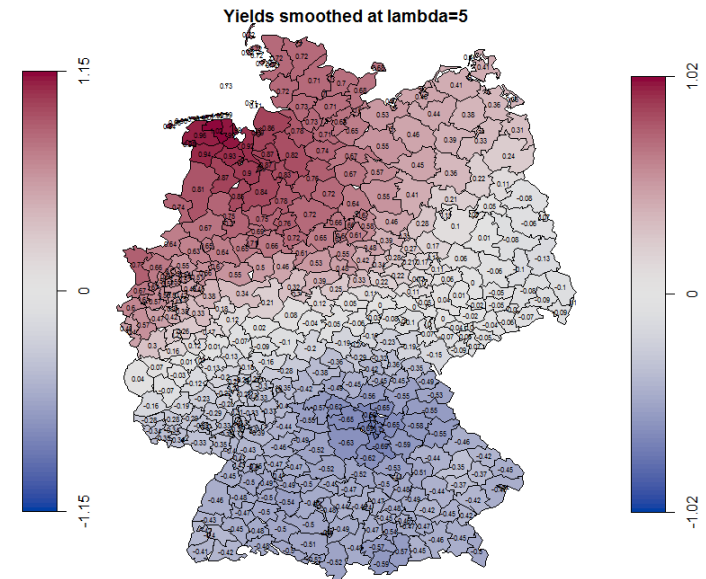
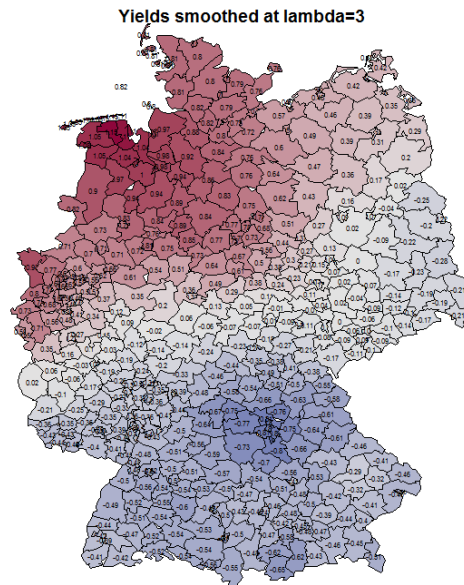
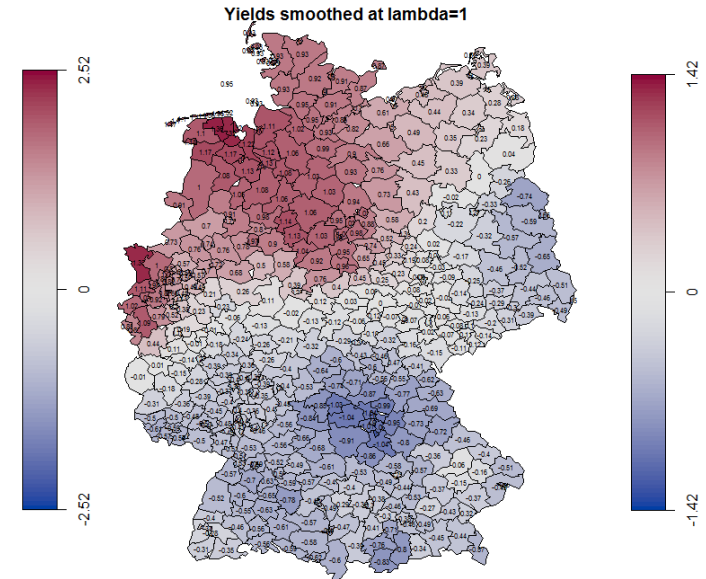
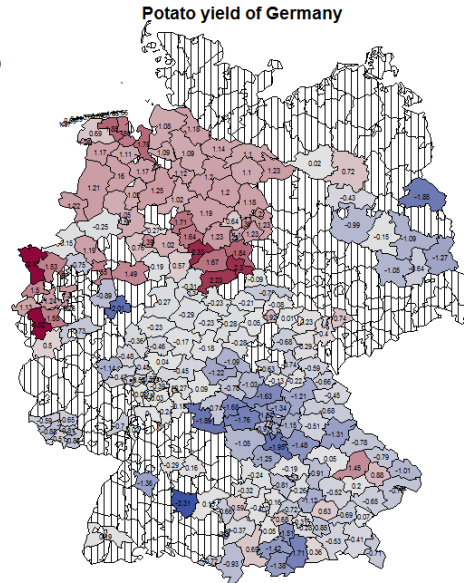
- Yield data for 230 out of 439 districts
- Estimation steps:
 - Penalty matrix K
 - Via neighborhood
 - Ruegen's neighbor
 - Nordvorpommern
 - Design matrix Z
 - Via yield data
 - PLS estimation
 - $\hat{\gamma} = (Z'Z + \lambda K)^{-1}Z'y$



Before		
	13057	13061
13057	5	0
13061	0	0
After		
	13057	13061
13057	6	-1
13061	-1	1

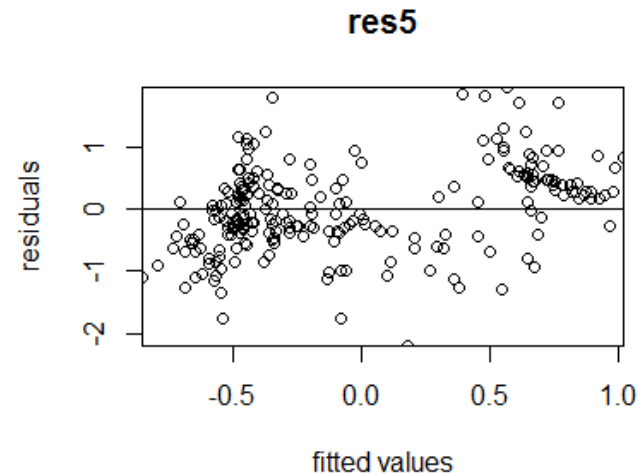
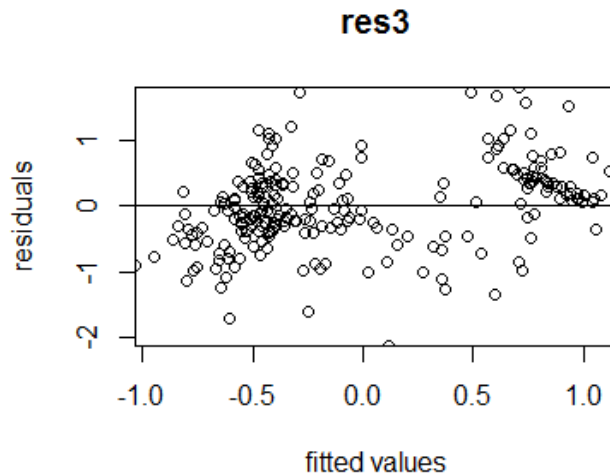
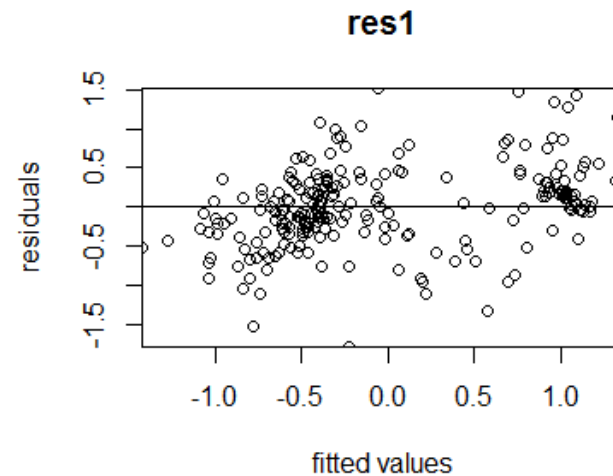
Results

- as $\lambda \uparrow$
 $smth \uparrow$
- but,
as $\lambda \uparrow$
 $\sigma_\varepsilon^2 \uparrow$



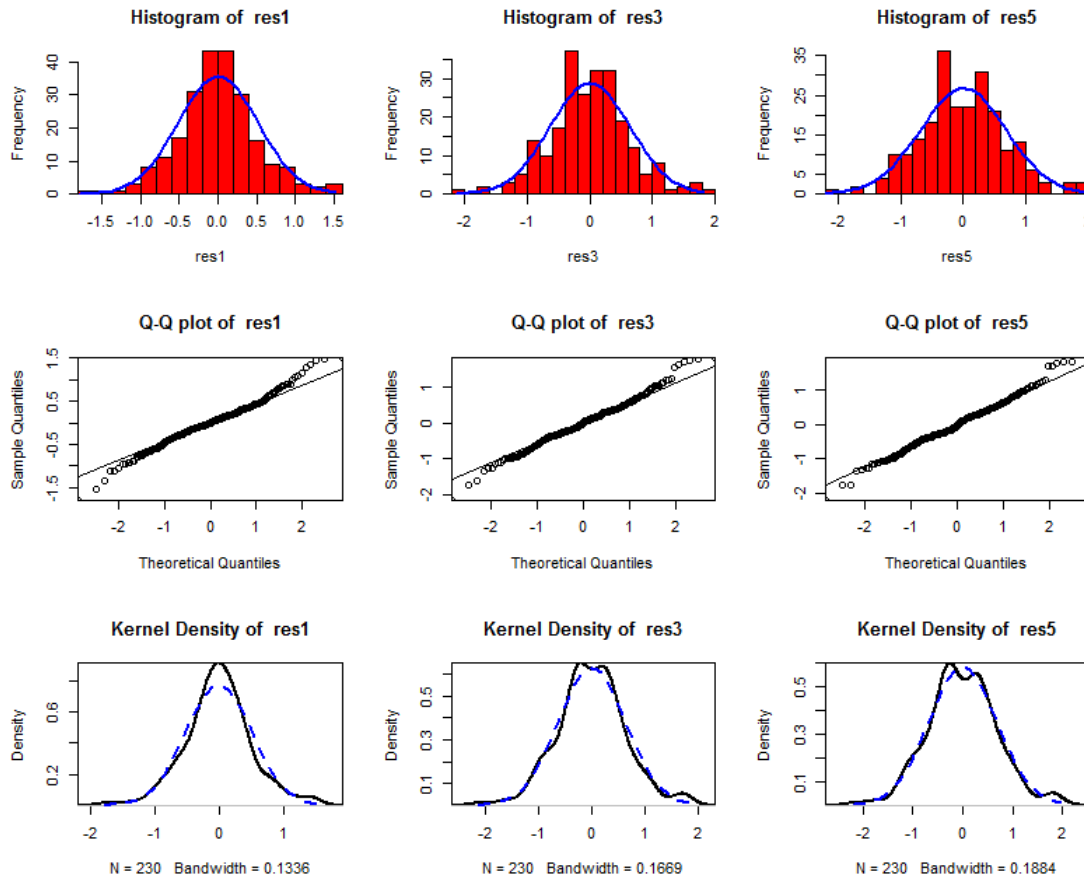
Assumptions

- \sim homoscedastic and linear residuals
- $\sigma_{\varepsilon|\lambda=1}^2 = 0.2673956$
- $\sigma_{\varepsilon|\lambda=3}^2 = 0.4040608$
- $\sigma_{\varepsilon|\lambda=5}^2 = 0.4691135$



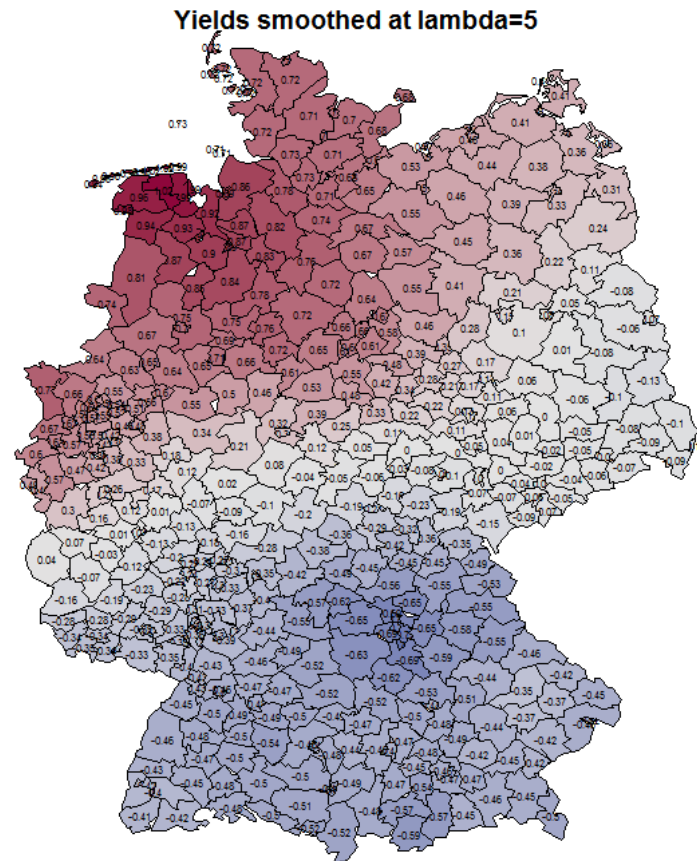
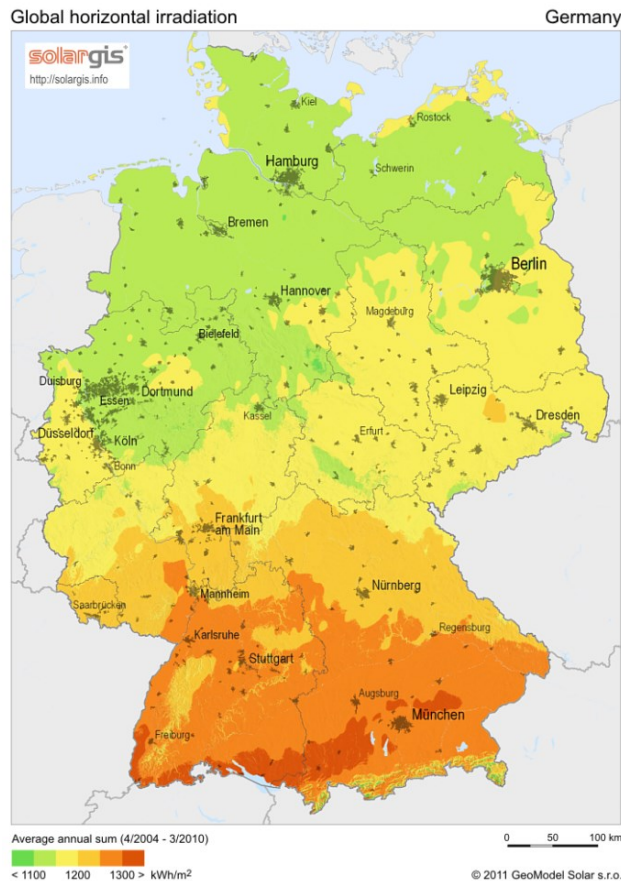
Assumptions contd.

- \sim normally distributed residuals



Reason

- Solar potential? (no, probably other gainful products)



Source: Wikipedia, 2017



Thank you...

Caner Erdem

Sources

- Fahrmeir, L., Kneib, T., Lang, S., & Marx, B. (2013). Regression: Models, Methods and Applications. Berlin Heidelberg: Springer-Verlag.
- Fleet, D., & Jepson, A. Markov Random Fields. Retrieved July 08, 2017, from <http://www.cs.toronto.edu/~kyros/courses/2503/Handouts/mrf.pdf>
- Wikipedia. Solar power in Germany. Retrieved July 12, 2017, from https://en.wikipedia.org/wiki/Solar_power_in_Germany
- eBay. 8 Reasons Why Playing in the Sand Is Good for Kids. Retrieved July 12, 2017, from <http://www.ebay.com/gds/8-Reasons-Why-Playing-in-the-Sand-Is-Good-for-Kids-/10000000177634049/g.html>