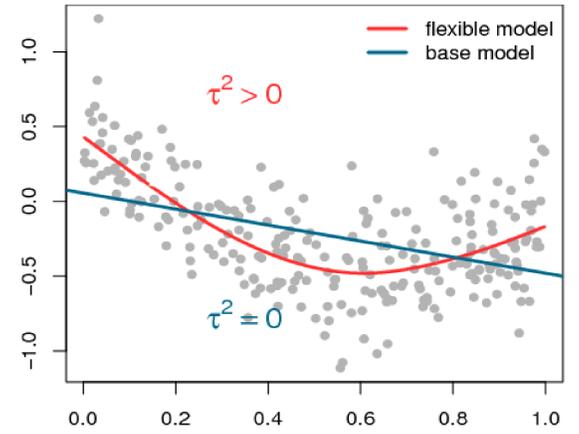


Base Models and Increased Complexity

- Nested structure inherent in many model components.
- Hyperparameter τ^2 to determine the deviation of a flexible alternative to the base model.
- **Aims:** Construct principled prior distributions for variances of additive components in structured additive distributional regression (motivated from penalised complexity priors²). Develop Bayesian inference and elaborate similar approach for other priors from the literature.



- Alternative priors for τ^2 include half-normal (HN), half-Cauchy (HC) and uniform (U) priors for τ .
- HN priors for τ correspond to gamma, HC to generalised beta priors for τ^2 .

Distributional Regression

M1: $y | x \sim F$ with parametric density $f(y | \vartheta_1, \dots, \vartheta_K)$

M2: $\vartheta_k(x) = h_k(\eta_k)$, $k = 1, \dots, K$ with structured additive predictors

$$\eta_k = \sum_j g_j(x) = \sum_j Z_j \beta_j$$

Standard Prior Specifications

- Regularisation or smoothness can be achieved via $p(\beta | \tau^2) \propto \exp(-1/(2\tau^2)\beta'K\beta)$ with prior precision K (specific to j, k , often rank deficient).
- For variances τ^2 inverse gamma (IG(a, b)) priors are used:

$a, b = \varepsilon$, small (flat on log-scale)
 $a = -1, b = 0$ (flat for τ^2)

$a = 1, b = \varepsilon$ (flat for $1/\tau^2$)
 $a = -0.5, b = 0$ (flat for τ).

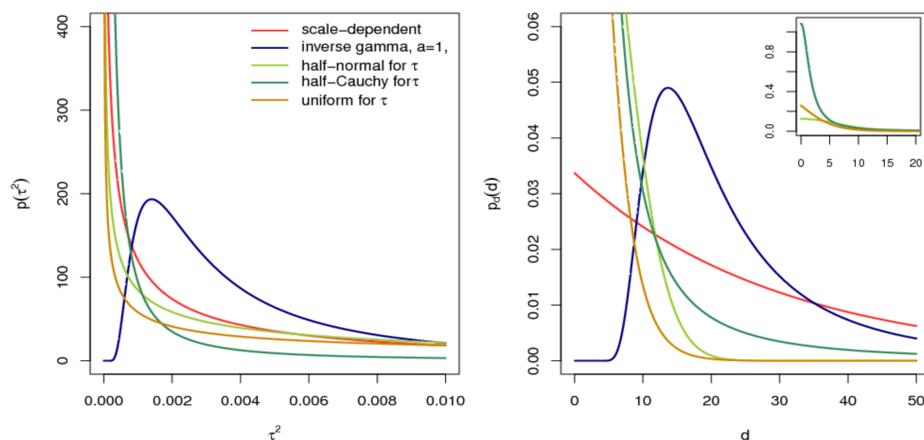
Scale-Dependent (SD) Priors

SD prior for τ^2 has density of a Weibull distribution with shape parameter $a=0.5$ and scale parameter θ resulting from three construction principles²:

Occam's razor: Prior should invoke principle of parsimony.

Measure of complexity: Increased complexity between base model p_b and flexible alternative p is measured by $d(p || p_b) = \sqrt{2\text{KLD}(p || p_b)}$.

Constant rate penalisation: Implies exponential prior on distance scale d , i.e. constant rate of decay in the distance prior from the base model to stronger deviations.



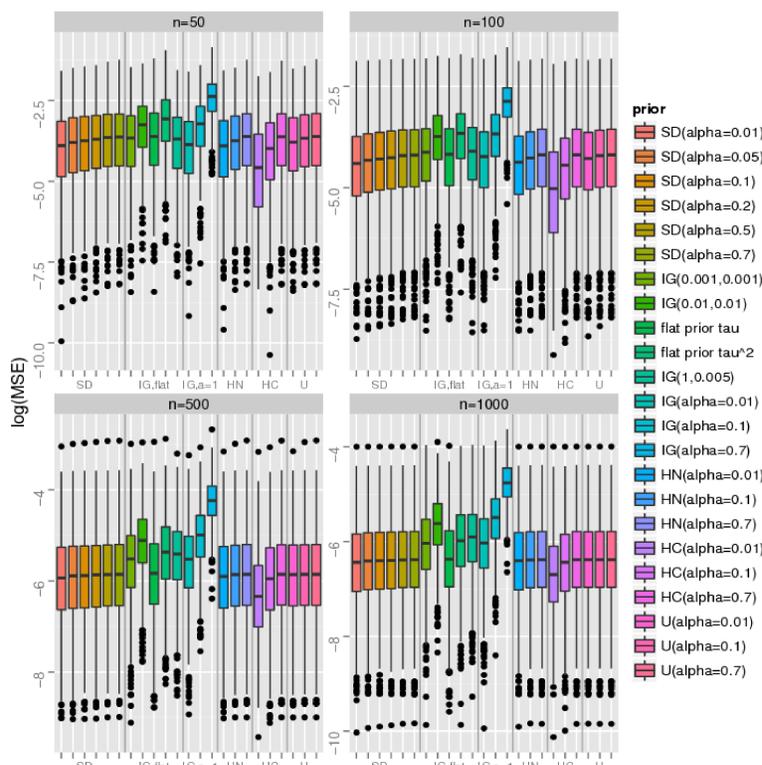
User-Defined Scaling

- Decay rate $\exp(-\lambda)$ in the exponential prior is controlled implicitly through scale parameter θ of the prior for τ^2 by $P(q(\tau^2 \leq c)) = 1 - \alpha$.
- Prior knowledge on effects $g = Z\beta$ allows to specify interval with high marginal probability, $P(|g(x)| \leq c \forall x \in D) \geq 1 - \alpha$. Resulting integral can be approximated numerically.
- User-defined scaling has been transferred to HN, HC and U priors for τ (and IG priors for τ^2 when $a=1$).
- Optimising θ for given Z and K is implemented in the R-package **sdPrior** on CRAN.

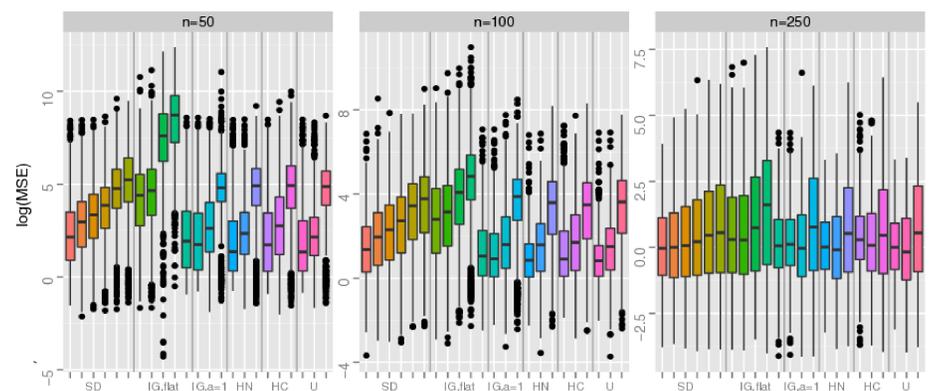
MCMC Inference

- Full conditional $p(\tau^2 | \cdot)$ has no closed form when $p(\tau^2)$ is SD prior or priors derived from HN, HC and U for τ .
- Still feasible with Metropolis-Hastings steps, similar as for β .
- Construction of proposal densities based on approximations to $\log(p(\log(\tau^2) | \cdot))$.
- Empirically, MCMC chains have good convergence behaviour.
- Conditions for propriety of joint posterior with SD prior are derived.
- Implementation in open software BayesX (www.bayesx.org).

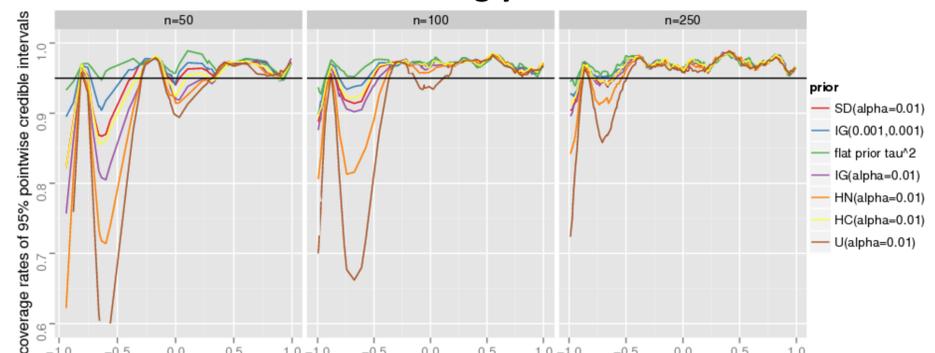
Simulation 1: Close to the Base Model



Simulation 2: Flat Likelihood



Simulation 3: Strongly Nonlinear Effect



Overall: SD priors are fairly robust in all considered scenarios.