Prior and posterior predictive checking

- Bayesian Data Analysis, 3rd ed, Chapter 6
- Jonah Gabry, Daniel Simpson, Aki Vehtari, Michael Betancourt, and Andrew Gelman (2018). Visualization in Bayesian workflow. Journal of the Royal Statistical Society Series A, accepted for publication as discussion paper. arXiv preprint arXiv:1709.01449.
- Graphical posterior predictive checks using the bayesplot package http://mc-stan.org/bayesplot/articles/graphical-ppcs.html
- demo demos_rstan/ppc/poisson-ppc.Rmd
- Workflow with prior and posterior predictive checking https://betanalpha.github.io/assets/case_studies/ principled_bayesian_workflow.html

Replicates vs. future observation

 Predictive ỹ is the next not yet observed possible observation. y^{rep} refers to replicating the whole experiment (with same values of x) and obtaining as many replicated observations as in the original data.

- Data y
- Parameters θ
- Replicated data y^{rep}
 - assume that the data has been generated by a process which can be well described by the model *M* with parameters θ
 - replicated data could be observed if the experiment were repeated
 - replace "true" data generating process by the model

$$p(y^{\mathrm{rep}}|y, M) = \int p(y^{\mathrm{rep}}|\theta, M) p(\theta|y, M) d\theta$$

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• Graphical checking: compare distribution plots

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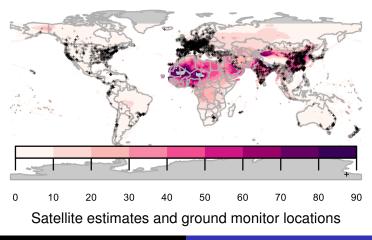
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- Graphical checking: compare distribution plots
- Test quantity (or discrepancy measure) $T(y, \theta)$
 - summary quantity used to compare the observed data and replicates from the predictive distribution

- Similar to posterior predictive checking but without data
- Draws from the prior predictive distribution are compared to external information

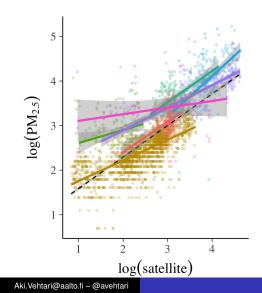
- Gabry et al (2017). Visualization in Bayesian workflow.
 - Estimation of human exposure to air pollution from particulate matter measuring less than 2.5 microns in diameter (PM_{2.5})
 - A recent report estimated that PM_{2.5} is responsible for three million deaths worldwide each year (Shaddick et al, 2017)

• Gabry et al (2017). Visualization in Bayesian workflow.

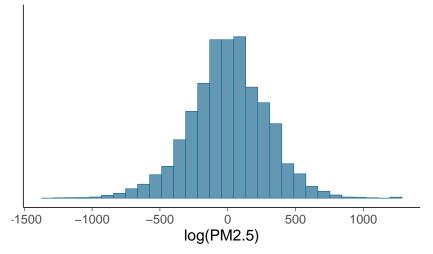


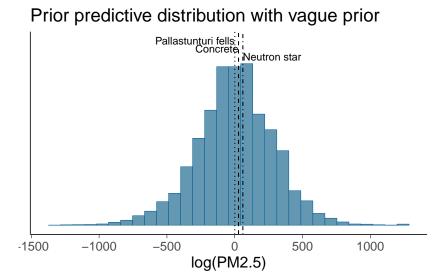
Aki.Vehtari@aalto.fi - @avehtari

• Gabry et al (2017). Visualization in Bayesian workflow.

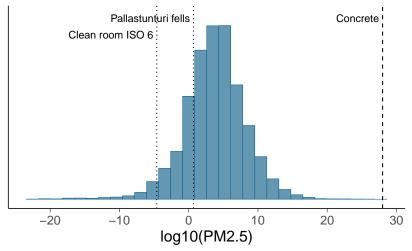


Prior predictive distribution with vague prior



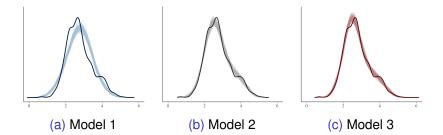


Prior predictive distribution with weakly informative

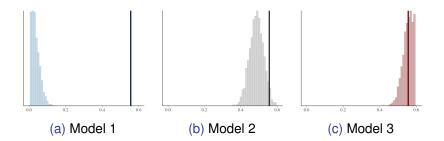


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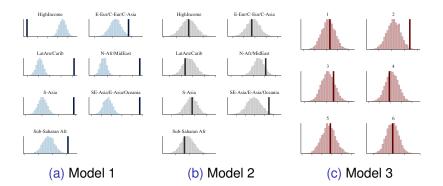
Posterior predictive checking - predictive distributions



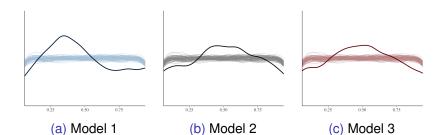
Posterior predictive checking – test statistic (skewness)



Posterior predictive checking - test statistic (median for groups)



LOO predictive checking - LOO-PIT



• demo demos_rstan/ppc/poisson-ppc.Rmd

```
data -
  int<lower=1> N:
  int < lower=0> y [N];
parameters {
  real<lower=0> lambda;
model {
  lambda ~ exponential (0.2);
  y poisson(lambda);
generated quantities {
  real log_lik[N];
  int y_rep[N];
  for (n in 1:N) {
    y_rep[n] = poisson_rng(lambda);
    \log_{lik}[n] = poisson_{lpmf}(y[n] | lambda);
```