

# Bayesian Learning

## Lecture 12 - Model evaluation

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- Model evaluation - Posterior predictive analysis

# Models - why?

- We now know how to **compare** models.
- But how do we know if any given model is 'any good'?
- George Box: '**All models are false, but some are useful**'.

# What is your model for?

## ■ Prediction.

- ▶ Interpretation not a concern
- ▶ Black-box approach may be ok.
- ▶ Model averaging may be a good idea.

## ■ Abstraction to aid in thinking about a phenomena.

- ▶ Prediction accuracy of less concern.
- ▶ Model averaging may be a bad idea.

## ■ Model as a compact description of a complex phenomena.

- ▶ Computational cost of model evaluation may be a concern.
- ▶ Online/real-time analysis.

# Posterior predictive analysis

- If  $p(y|\theta)$  is a 'good' model, then the data actually observed should not differ 'too much' from simulated data from  $p(y|\theta)$ .
- Bayesian: simulate data from the **posterior predictive distribution**:

$$p(y^{rep}|y) = \int p(y^{rep}|\theta)p(\theta|y)d\theta.$$

- Difficult to compare  $y$  and  $y^{rep}$  because of dimensionality.
- Solution: compare **low-dimensional statistic**  $T(y, \theta)$  to  $T(y^{rep}, \theta)$ .
- Evaluates the full probability model consisting of both the likelihood *and* prior distribution.

# Posterior predictive analysis

- **Algorithm** for simulating from the posterior predictive density  $p[T(y^{rep})|y]$ :
  - 1 Draw a  $\theta^{(1)}$  from the posterior  $p(\theta|y)$ .
  - 2 Simulate a data-replicate  $y^{(1)}$  from  $p(y^{rep}|\theta^{(1)})$ .
  - 3 Compute  $T(y^{(1)})$ .
  - 4 Repeat steps 1-3 a large number of times to obtain a sample from  $T(y^{rep})$ .
- We may now compare the observed statistic  $T(y)$  with the distribution of  $T(y^{rep})$ .
- **Posterior predictive p-value:**  $\Pr[T(y^{rep}) \geq T(y)]$
- Informal graphical analysis.

# Posterior predictive analysis - Examples

- Ex. 1. Model:  $y_1, \dots, y_n | \mu, \sigma^2 \stackrel{iid}{\sim} N(\mu, \sigma^2)$ .  $T(y) = \max_i |y_i|$ .
- Ex. 2. ARIMA-process.  $T(y)$  may be the autocorrelation function.
- Ex. 3. Poisson regression.  $T(y)$  frequency distribution of the response counts. Proportions of zero counts.

# Posterior predictive analysis - Normal model, max statistic

