

## Previously on MA40189:

- Markov Chain Monte Carlo:
  - construct a Markov chain that has the required posterior distribution as its stationary distribution
  - sample from the chain until (approx) convergence
  - from this point, samples from the chain are viewed as a sample of values from the posterior distribution
  - use the resulting sample (from the stationary distribution) to estimate properties of the posterior distribution

## Today on MA40189:

- each irreducible and aperiodic Markov chain  $X_t$  has a unique stationary distribution  $\pi(x)$  which it converges to
- ergodic (aperiodic and positive recurrent) Markov chain

$$P \left\{ \frac{1}{N} \sum_{i=1}^N g(X_i) \rightarrow E\{g(X) \mid X \sim \pi(x)\} \right\} = 1.$$

- i.e. ergodic averages converge to desired expectations
- we want to draw samples from  $f(\theta \mid x) = cg(\theta)$ 
  - strategy is to set-up an irreducible and aperiodic Markov chain whose stationary distribution  $\pi(\theta) = f(\theta \mid x)$
  - achieve this using the Metropolis-Hastings algorithm