

## Previously on MA40189:

- pdf of exchangeable  $X_1, X_2, \dots$  can be expressed as

$$f(x_1, \dots, x_n) = \int_{\theta} \left\{ \prod_{i=1}^n f(x_i | \theta) \right\} f(\theta) d\theta$$

- can view the  $X_i$  as **conditionally independent** given  $\theta$
- $t(X)$  **sufficient** for  $X$  for learning about  $\theta$  if

$$f(x | \theta) = g(t, \theta)h(x)$$

- **$k$ -parameter exponential** family:

$$f(x | \theta) = \exp \left\{ \sum_{j=1}^k \phi_j(\theta) u_j(x) + g(\theta) + h(x) \right\}$$

## Today on MA40189:

- **exchangeable sequences** with exponential family likelihood have **sufficient statistics**:

$$t_n = t_n(X_1, \dots, X_n) = \left[ n, \sum_{i=1}^n u_1(X_i), \dots, \sum_{i=1}^n u_k(X_i) \right]$$

- **conjugacy** of exponential family distributions:

- regard  $f(x | \theta)$  as a function of  $\theta$ ,

$$f(x | \theta) = \exp \left\{ \sum_{j=1}^k u_j(x) \phi_j(\theta) + g(\theta) + h(x) \right\}$$

- posterior is

$$f(\theta | x) \propto \exp \left\{ \sum_{j=1}^k (a_j + u_j(x)) \phi_j(\theta) + (d + 1)g(\theta) \right\}$$