# 1 Probability distributions

- 1.1 Beta distribution
- 1.2 Binomial distribution

# 2 Bayesian parameter estimation

# 2.1 Beta-Bernoulli model

#### **2.1.1 Summary**

The model

$$X_i \sim \text{Ber}(\theta), \text{ for } i \in \{1, \dots, N\}$$
 (2.1)

$$\mathcal{D} = \{x_1, \dots, x_N\} \tag{2.2}$$

$$N_1 = \sum_{i=1}^{N} \mathbb{I}(x_i = 1) \tag{2.3}$$

$$N_0 = \sum_{i=1}^{N} \mathbb{I}(x_i = 0) \tag{2.4}$$

Likelihood

$$p(\mathcal{D}|\theta) = \theta^{N_1} (1 - \theta)^{N_0} \tag{2.5}$$

**Prior** 

$$p(\theta) = \text{Beta}(\theta|a, b)$$
 (2.6)

**Posterior** 

$$p(\theta|\mathcal{D}) = \text{Beta}(\theta|N_1 + a, N_0 + b) \tag{2.7}$$

Posterior predictive

$$p(x|\mathcal{D}) = \frac{a}{a+b} \tag{2.8}$$

#### 2.1.2 Derivations

#### 2.2 Beta-binomial model

## 2.2.1 Summary

The model

$$N_1 \sim \text{Bin}(N, \theta)$$
 (2.9)

$$\mathcal{D} = \{N_1, N\} \tag{2.10}$$

$$N_1 = \text{number of successes}$$
 (2.11)

$$N = \text{total number of trials}$$
 (2.12)

$$\mathcal{D}' = \{N_1', N'\} \tag{2.13}$$

$$N_1' = \text{number of successes in a new batch of data}$$
 (2.14)

$$N' = \text{total number of trials in a new batch of data}$$
 (2.15)

Likelihood

$$p(\mathcal{D}|\theta) = \operatorname{Bin}(N_1|N,\theta) \tag{2.16}$$

**Prior** 

$$p(\theta) = \text{Beta}(\theta|a, b) \tag{2.17}$$

**Posterior** 

$$p(\theta|\mathcal{D}) = \text{Beta}(\theta|N_1 + a, N_0 + b)$$
(2.18)

Posterior predictive

$$p(\mathcal{D}'|\mathcal{D}) = Bb(N_1'; a, b, N') \tag{2.19}$$

#### 2.2.2 Derivations

# 2.3 Dirichlet-categorical model

#### **2.3.1 Summary**

The model

$$X_i \sim \operatorname{Cat}\left(\boldsymbol{\theta} = (\theta_1, \dots, \theta_K)^T\right), \text{ for } i \in \{1, \dots, N\}$$
 (2.20)

$$\mathcal{D} = \{x_1, \dots, x_N\} \tag{2.21}$$

$$N_k = \sum_{i=1}^{N} \mathbb{I}(x_i = k)$$
 (2.22)

Likelihood

$$p(\mathcal{D}|\theta) = \prod_{k=1}^{K} \theta_k^{N_k}$$
 (2.23)

Prior

$$p(\theta) = \text{Dir}(\theta; \alpha) \tag{2.24}$$

**Posterior** 

$$p(\theta|\mathcal{D}) = \text{Dir}(\boldsymbol{\theta}; \alpha_1 + N_1, \dots, \alpha_K + N_K)$$
 (2.25)

Posterior predictive

$$p(X = j|\mathcal{D}) = \frac{\alpha_j + N_j}{\alpha_0 + N}$$
 (2.26)

$$p(X = j | \mathcal{D}) = \frac{\alpha_j + N_j}{\alpha_0 + N}$$
where  $\alpha_0 = \sum_{k=1}^K \alpha_k$  (2.26)

### 2.3.2 Derivations