

# Hand Calculation Example: Beta-Binomial Model

## Problem Setup

We observe  $y = 1$  success in  $N = 7$  trials. We use a uniform prior ( $a_0 = 1, b_0 = 1$ ).

## Step 1: Prior Distribution

Prior:  $\text{Beta}(1, 1) = \text{Uniform}(0, 1)$

- Prior mean:  $E[\theta] = a_0/(a_0+b_0) = 1/(1+1) = 0.5$
- Prior variance:  $V[\theta] = (a_0 b_0)/[(a_0+b_0)^2(a_0+b_0+1)] = (1 \times 1)/[(2)^2(3)] = 1/12 \approx 0.083$

## Step 2: Likelihood

Likelihood:  $p(y=1|\theta) = C(7,1) \times \theta^1 \times (1-\theta)^6 = 7\theta(1-\theta)^6$

## Step 3: Posterior Distribution

Using the conjugacy property:

- Posterior:  $\text{Beta}(a_0 + y, b_0 + N - y) = \text{Beta}(1 + 1, 1 + 7 - 1) = \text{Beta}(2, 7)$

## Posterior Statistics:

- **Posterior mean:**  $E[\theta|y] = 2/(2+7) = 2/9 \approx 0.222$
- **Posterior mode:**  $\text{Mode}[\theta|y] = (2-1)/(2+7-2) = 1/7 \approx 0.143$
- **Posterior variance:**  $V[\theta|y] = (2 \times 7)/[(9)^2(10)] = 14/810 \approx 0.017$

## For comparison:

- **MLE:**  $\theta_{\text{MLE}} = y/N = 1/7 \approx 0.143$
- **95% Credible Interval:** Use  $\text{Beta}(2,7)$  quantiles
  - Lower:  $\text{Beta}(2,7).\text{ppf}(0.025) \approx 0.028$
  - Upper:  $\text{Beta}(2,7).\text{ppf}(0.975) \approx 0.503$

## Step 4: Posterior Predictive

For  $N^* = 5$  new trials, the posterior predictive is Beta-Binomial(5, 2, 7):

$$P(y^* = k | y) = C(5,k) \times B(k+2, 5-k+7) / B(2,7)$$

Where  $B(a,b)$  is the Beta function.

Example calculations:

- $P(y^* = 0 \mid y) \approx 0.318$
- $P(y^* = 1 \mid y) \approx 0.409$
- $P(y^* = 2 \mid y) \approx 0.205$
- $E[y^* \mid y] = N^* \times E[\theta \mid y] = 5 \times (2/9) \approx 1.11$

## Step 5: Model Evidence

$$p(y) = C(7,1) \times B(1+1, 7-1+1) / B(1,1) = 7 \times B(2,7) / B(1,1) = 7 \times [\Gamma(2)\Gamma(7)/\Gamma(9)] / [\Gamma(1)\Gamma(1)/\Gamma(2)] = 7 \times [1! \times 6!/8!] / [1] = 7 \times (6!)/(8 \times 7 \times 6!) = 7 \times 1/(8 \times 7) = 1/8 = 0.125$$

## Sequential Update Example

Starting with uniform prior Beta(1,1):

1. Observe ( $N_1=5, y_1=2$ ):
  - Update to Beta(1+2, 1+5-2) = Beta(3, 4)
  - Mean:  $3/7 \approx 0.429$
2. Observe ( $N_2=3, y_2=1$ ):
  - Update to Beta(3+1, 4+3-1) = Beta(4, 6)
  - Mean:  $4/10 = 0.400$
3. Observe ( $N_3=4, y_3=3$ ):
  - Update to Beta(4+3, 6+4-3) = Beta(7, 7)
  - Mean:  $7/14 = 0.500$

Final posterior: Beta(7, 7)

Verification: Total data is  $N=12, y=6$  Direct calculation: Beta(1+6, 1+12-6) = Beta(7, 7) ✓

## Key Insights

1. **Effect of prior:** With small data ( $N=7, y=1$ ), the uniform prior pulls the posterior mean (0.222) away from the MLE (0.143) toward 0.5.
2. **Shrinkage:** The posterior mean is a weighted average between the prior mean and the MLE, with weights determined by the "strength" of prior and data.
3. **Uncertainty:** The posterior variance (0.017) is much smaller than the prior variance (0.083), showing how data reduces uncertainty.

4. **Conjugacy benefit:** The Beta-Binomial conjugacy allows all calculations to be done analytically without numerical integration.