



Course Notes Summary

Focus Area: Summary of main formulas please

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Prepared for quick revision and reference

Use this sheet as a step-by-step guide when solving problems.

1 Key definitions and concepts

- Conditional probability: $P(B | A)$ is the probability B occurs given A has occurred.
- Prior probability: initial probability before new information.
- Posterior probability: revised probability after incorporating new information.
- Likelihood: $P(\text{new data} | \text{hypothesis})$, used to update the prior.
- Joint probability: $P(A \cap B)$ is the probability both A and B occur.
- Law of total probability: expresses $P(A)$ via a partition of the sample space.

2 Fundamental formulas

- Conditional probability (definition)

$$P(B | A) = \frac{P(A \cap B)}{P(A)}$$

- Symmetric expression for joint probability

$$P(A \cap B) = P(A) P(B | A) = P(B) P(A | B)$$

- Bayes' theorem (basic two-event form)

$$P(B | A) = \frac{P(A | B) P(B)}{P(A)}$$

- Law of total probability (for a partition B_1, \dots, B_n)

$$P(A) = \sum_{i=1}^n P(A | B_i) P(B_i)$$

- Bayes' theorem (generalized for a partition B_1, \dots, B_n)

$$P(B_j | A) = \frac{P(A | B_j) P(B_j)}{\sum_{i=1}^n P(A | B_i) P(B_i)}$$

3 Important relationships and notes

- Prior \rightarrow Likelihood \rightarrow Posterior sequence:
 1. Start with prior $P(B_j)$.
 2. Use likelihood $P(A | B_j)$.
 3. Compute posterior $P(B_j | A)$ via Bayes' theorem.
- Normalizing constant in Bayes' rule is $P(A)$ given by the law of total probability.
- $P(B | A)$ increases when $P(A | B)$ is relatively large compared with $P(A | \text{other events})$ weighted by priors.
- Bayes can be applied with discrete counts or probabilities; both formulations are equivalent.

4 Intuitive (frequency/table) method

- Method:

1. Assume a convenient total N (e.g., 1000 or 100000).
2. Convert probabilities to counts: $\text{count}(B_i) = N \cdot P(B_i)$.
3. Compute cells: $\text{count}(A \cap B_i) = \text{count}(B_i) \cdot P(A | B_i)$.
4. Use counts to compute desired conditional:

$$P(B_j | A) = \frac{\text{count}(A \cap B_j)}{\sum_i \text{count}(A \cap B_i)}.$$

- Advantage: reduces algebra errors, makes denominators explicit.

5 Worked examples

- Example A: Orange County cigar smokers

- Given:

- * $P(\text{male}) = 0.51$
- * $P(\text{female}) = 0.49$
- * $P(\text{cigar} | \text{male}) = 0.095$
- * $P(\text{cigar} | \text{female}) = 0.017$

- Bayes calculation:

$$\begin{aligned} P(\text{male} | \text{cigar}) &= \frac{P(\text{cigar} | \text{male}) P(\text{male})}{P(\text{cigar} | \text{male}) P(\text{male}) + P(\text{cigar} | \text{female}) P(\text{female})} = \frac{0.095 \cdot 0.51}{0.095 \cdot 0.51 + 0.017 \cdot 0.49} \\ &= \frac{0.04845}{0.04845 + 0.00833} \approx \frac{0.04845}{0.05678} \approx 0.853 \end{aligned}$$

- Frequency/table check ($N = 100,000$):

- * males = 51,000; male cigar = $0.095 \times 51,000 = 4,845$
- * females = 49,000; female cigar = $0.017 \times 49,000 = 833$
- * total cigar = $4,845 + 833 = 5,678$
- * $P(\text{male} | \text{cigar}) = \frac{4,845}{5,678} \approx 0.853$

- Example B: ELT manufacturers and defectives

- Given:

- * $P(A) = 0.80$, $P(B) = 0.15$, $P(C) = 0.05$
- * $P(\text{defective} | A) = 0.04$, $P(\text{defective} | B) = 0.06$, $P(\text{defective} | C) = 0.09$

- Bayes calculation:

$$\begin{aligned} P(A | \text{defective}) &= \frac{P(\text{defective} | A) P(A)}{P(\text{defective} | A) P(A) + P(\text{defective} | B) P(B) + P(\text{defective} | C) P(C)} = \frac{0.032}{0.032 + 0.009 + 0.0045} = \frac{0.032}{0.0455} \approx 0.703 \end{aligned}$$

- Frequency/table check ($N = 10,000$):

- * A total = 8,000; defective from $A = 0.04 \times 8,000 = 320$
- * B total = 1,500; defective from $B = 0.06 \times 1,500 = 90$
- * C total = 500; defective from $C = 0.09 \times 500 = 45$
- * total defective = $320 + 90 + 45 = 455$
- * $P(A | \text{defective}) = \frac{320}{455} \approx 0.703$

6 Quick solution checklist (ordered steps)

1. Identify hypothesis events B_1, \dots, B_n and the observed event A .
2. Record priors $P(B_i)$ and likelihoods $P(A | B_i)$.
3. Compute $P(A) = \sum_i P(A | B_i) P(B_i)$.
4. Compute posterior for target B_j :

$$P(B_j | A) = \frac{P(A | B_j) P(B_j)}{P(A)}.$$

5. Optionally verify with frequency/table method using a convenient N .

7 Compact reference formulas (for quick revision)

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$$P(B | A) = \frac{P(A | B) P(B)}{P(A)}$$

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$$P(A) = \sum_i P(A | B_i) P(B_i)$$

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$$P(A \cap B) = P(A) P(B | A) = P(B) P(A | B)$$