

Conditional Probability

- **Q:** Should the following probabilities be different?

➤ Event = 王建民本季戰績至少獲得6次勝投

- $P=??$ in the beginning of the season
- $P=??$ in the middle of the season
- $P=??$ now

➤ Event = rain tomorrow

- $P=??$ if no information about where you are staying
- $P=??$ if you are staying in a desert
- $P=??$ if a typhoon will hit the place you stay tomorrow

- **Q:** What causes the differences?

➤ For an event, *new information* (i.e., some other event has occurred) could change its probability

➤ We call the altered probability a conditional probability

- Mathematical Definition: If A and B are two events in a sample space Ω and $P(A)>0$, then

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

is called the *conditional probability* of B given A .

➤ In the classical model,

$$P(A) = \#A/\#\Omega \quad \text{and} \quad P(A \cap B) = \#(A \cap B)/\#\Omega$$

$$\Rightarrow P(B|A) = \frac{\#(A \cap B)/\#\Omega}{\#A/\#\Omega} = \frac{\#(A \cap B)}{\#A}$$

- Example: A family is known to have 2 children, *at least one of whom is a girl*. **Q:** Probability that the other is a boy=??

$$\square \Omega = \{bb, bg, gb, gg\}$$

$$\square A = \{bg, gb, gg\} \quad \text{and} \quad B = \{bb, bg, gb\}$$

$$\square P(B|A) = \#(A \cap B)/\#A = 2/3$$

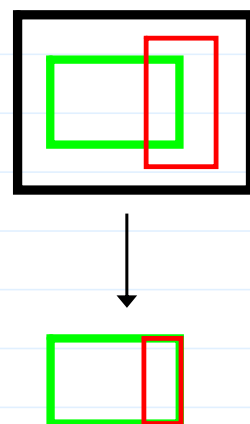
- Note: $\#\Omega$ is reduced to $\#A$.

➤ In effect by conditioning,

- we are restricting the sample space from Ω to A , i.e., $\Omega \rightarrow A$,

- and, for an arbitrary event B in Ω to occur when A has occurred, we need that both A and B occur together, i.e., $B \rightarrow B \cap A$.

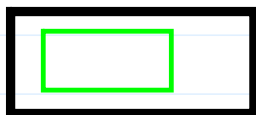
➤ The division by $P(A)$ in the definition above rescales all probabilities from the entire sample space Ω to being relative to the new sample space A



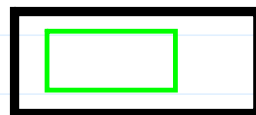
➤ $P(B|A)$ is a probability measure defined on B , but not A .

- $P(\cdot|A)$ satisfies the 3 axioms of probability (**exercise**)

$P(\cdot)$



$P(\cdot|A)$



- Any propositions developed in Chapter 2 for probability measures can be applied on $P(\cdot|A)$.
(For example, $P(B^c|A) = 1 - P(B|A)$.)

• 3 Useful Formulas for Calculating Probabilities
(for 2-events case)

1. If $P(A) > 0$, then $P(A \cap B) = P(A)P(B|A)$.



2. If $0 < P(A) < 1$, then

$$P(B) = P(A)P(B|A) + P(A^c)P(B|A^c).$$



3. If $0 < P(A) < 1$ and $P(B) > 0$, then

$$P(A|B) = \frac{P(A)P(B|A)}{P(A)P(B|A) + P(A^c)P(B|A^c)}.$$



➤ Example (Urn Problem)

- The Story. n balls sequentially and randomly chosen, *without replacement*, from an urn containing R red and $N-R$ white balls ($n \leq N$). **Q**: Given that k of the n balls are red ($k \leq R$), probability that the 1st ball chosen is red = ??

- Let $A = \{k \text{ of the } n \text{ balls are red}\}$
 $B = \{1^{\text{st}} \text{ ball chosen is red}\}$

- Method 1: $P(B|A) = \frac{\binom{n-1}{k-1}}{\binom{n}{k}} = k/n$

- Method 2:

$$\square P(A) = \left[\binom{R}{k} \times \binom{N-R}{n-k} \right] / \binom{N}{n}$$

$$\square P(A \cap B) = P(B)P(A|B) = \frac{R}{N} \times \frac{\binom{R-1}{k-1} \times \binom{N-R}{n-k}}{\binom{N-1}{n-1}}$$

$$\square P(B|A) = P(A \cap B) / P(A) = k/n$$

➤ Example (Diagnostic Tests)

- The Story. A diagnostic test for a *rare* disease (e.g., an Xray for lung cancer) is part of a routine physical exam.
- Let $\Omega = \{\text{the whole population of Taiwan}\}$
 $D = \{\text{disease is present}\}$



$E = \{\text{test indicates disease present}\}$

- Suppose that $P(D) = 0.001$, $P(E|D) = 0.98$, $P(E|D^c) = 0.01$

Q: Do you think the test is effective?

- Let us examine it from an alternative viewpoint

$$\begin{aligned} P(E) &= P(D)P(E|D) + P(D^c)P(E|D^c) \\ &= 0.001 \times 0.98 + 0.999 \times 0.01 = 0.01097. \end{aligned}$$

$$P(D|E) = \frac{P(D)P(E|D)}{P(D)P(E|D) + P(D^c)P(E|D^c)} = \frac{0.00098}{0.01097} = 0.0893.$$

$$P(D^c|E) = 1 - P(D|E) = 0.9107$$

Q: Now, do you still think the test is effective?

Q: But, why the 2 interpretations so different? What causes it?

- The probability of D increased by a factor of roughly 90 ($0.001 \rightarrow 0.0893$) when E occurs, but 0.0893 is still **small**

- The $P(E|D^c)$ ($=0.01$) and $P(E^c|D)$ ($=1 - P(E|D) = 0.02$) are called the *false positive* and *false negative rates*, respectively.

p. 3-6

➤ Example (Sampling Experiments): An urn contains R red balls and $N - R$ white balls. Sample 2 balls from the urn.

- $A = \{\text{red on the first draw}\}$
 $B = \{\text{red on the second draw}\}$



- Sampling *Without Replacement*:

$$P(A) = \frac{R(N-1)}{N(N-1)} = \frac{R}{N}, \quad P(A \cap B) = \frac{R(R-1)}{N(N-1)},$$

$$P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{(R(R-1))/(N(N-1))}{R/N} = \frac{R-1}{N-1}.$$

Similarly, $P(B|A^c) = \frac{R}{N-1}$. (**exercise**)

$$\begin{aligned} P(B) &= P(A)P(B|A) + P(A^c)P(B|A^c) \\ &= \frac{R}{N} \cdot \frac{R-1}{N-1} + \frac{N-R}{N} \cdot \frac{R}{N-1} \\ &= \frac{R^2 - R + NR - R^2}{N(N-1)} = \frac{R(N-1)}{N(N-1)} = \frac{R}{N}. \end{aligned}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{(R(R-1))/(N(N-1))}{R/N} = \frac{R-1}{N-1}.$$

□ Notes:

◆ The probabilities are proportional to # of red balls left

◆ $P(A|B) = P(B|A) \Rightarrow$ Symmetry.

■ Sampling With Replacement:

$$P(A) = \frac{R}{N} \cdot \frac{N}{N} = \frac{R}{N}, \quad P(B) = \frac{N}{N} \cdot \frac{R}{N} = \frac{R}{N},$$

$$P(A \cap B) = \frac{R}{N} \cdot \frac{R}{N} = \frac{R^2}{N^2}, \quad P(B|A) = \frac{R^2/N^2}{R/N} = \frac{R}{N} = P(B).$$

• Extensions of the 3 Useful Formulas (for m -events case)

1. (Multiplication Law) If A_1, \dots, A_m are events for which

$P(A_1 \cap \dots \cap A_{m-1}) > 0$, then

$$\begin{aligned} & P(A_1 \cap \dots \cap A_m) \\ &= P(A_1)P(A_2|A_1)P(A_3|A_1 \cap A_2) \cdots P(A_m|A_1 \cap \dots \cap A_{m-1}). \end{aligned}$$

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■ Example (Birthday Problem, LNp.2-3). Let A_k be the event that the k^{th} birthday differs from the first $k-1$. Then, $P(A_1)=1$,

$$\square P(A_k|A_1 \cap \dots \cap A_{k-1}) = \frac{365 - k + 1}{365}$$

$$\begin{aligned} \square P(A_1 \cap \dots \cap A_n) &= P(A_1)P(A_2|A_1) \cdots P(A_n|A_1 \cap \dots \cap A_{n-1}) \\ &= \prod_{k=1}^n \frac{365 - k + 1}{365} = \frac{365!}{365^n \cdot (365 - n)!} = \frac{(365)_n}{365^n}. \end{aligned}$$

■ Example (Matching Problem, LNp.2-11). **Q:** Probability that exactly k of n members have matches = ??

□ Let Ω be all permutations $\omega = (i_1, \dots, i_n)$ of $1, 2, \dots, n$.

□ Let $A_j = \{\omega: i_j = j\}$ and

$$A = \bigcup_{1 \leq j_1 < \dots < j_k \leq n} (A_1^c \cap \dots \cap A_{j_1-1}^c \cap A_{j_1} \cap A_{j_1+1}^c \cap \dots \cap A_n^c)$$

□ By symmetry, $P(A) = \binom{n}{k} \times P(A_1 \cap \dots \cap A_k \cap A_{k+1}^c \cap \dots \cap A_n^c)$

□ Let $E = A_1 \cap \dots \cap A_k$ and $G = A_{k+1}^c \cap \dots \cap A_n^c$

□ Then, $P(E \cap G) = P(E)P(G|E)$, where

$$P(E) = P(A_1)P(A_2|A_1) \cdots P(A_k|A_1 \cap \cdots \cap A_{k-1})$$

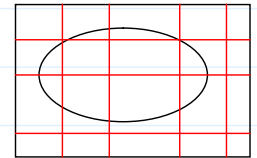
$$= \frac{1}{n} \times \frac{1}{n-1} \times \cdots \times \frac{1}{n-k+1} = \frac{(n-k)!}{n!} = \frac{1}{(n)_k}$$

$$\text{and, } P(G|E) = \sum_{i=0}^{n-k} (-1)^i \frac{1}{i!} \equiv p_{n-k}$$

$$\square P(A) = \binom{n}{k} \frac{1}{(n)_k} p_{n-k} = \frac{p_{n-k}}{k!} \approx \frac{e^{-1}}{k!}, \text{ when } n \text{ is large}$$

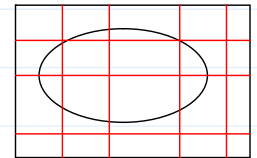
2.(Law of Total Probability) Let A_1, \dots, A_m be a partition of Ω and $P(A_i) > 0, i=1, \dots, m$, then for any event $B \subset \Omega$,

$$P(B) = \sum_{i=1}^m P(A_i)P(B|A_i).$$



3.(Bayes' Rule) Let A_1, \dots, A_m be a partition of Ω and $P(A_i) > 0, i=1, \dots, m$. If B is an event such that $P(B) > 0$, then for $1 \leq j \leq m$,

$$P(A_j|B) = \frac{P(A_j)P(B|A_j)}{\sum_{i=1}^m P(A_i)P(B|A_i)}.$$



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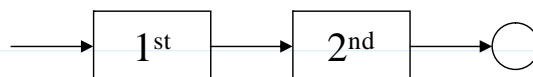
■ From Bayesians' viewpoint,

$P(A_j)$ = probability of A_j before B occurs \rightarrow prior prob.

$P(A_j|B)$ = probability of A_j after B occurs \rightarrow posterior prob.

\Rightarrow The Bayes' rule tells how to update the probabilities of A_j in light of the new information (i.e., B occurs)

■ An Application of Bayes' Rule. Suppose that a random experiment consists of two random stages



■ The probabilities of the 2nd-stage results depend on what happened in the 1st stage

■ We never see the result of the 1st stage, only the final result

■ We may be interested in finding the probability for outcomes in the 1st stage given the final result

➤ Example (Gold Coins):

■ The Story.

- ▣ Box 1 contains 2 silver coins.
- ▣ Box 2 contains 1 gold and 1 silver coin.
- ▣ Box 3 contains 2 gold coins.
- ▣ Experiment: (i) Select a box at random and, (ii) Examine the 2 coins in order (assuming all choices are equally likely at each stage)

■ **Q:** Given that 1st coin is gold, what is the probability that Box k is selected, $k=1, 2, 3$?

- ▣ Let $A_k = \{\text{Box } k \text{ is selected}\}$, $B = \{\text{1st coin is gold}\}$,

- ▣
$$P(B|A_k) = \begin{cases} 0, & \text{if } k = 1, \\ 1/2, & \text{if } k = 2, \\ 1, & \text{if } k = 3. \end{cases}$$

$$P(B) = \frac{1}{3} \cdot 0 + \frac{1}{3} \cdot \frac{1}{2} + \frac{1}{3} \cdot 1 = \frac{1}{2}.$$

- ▣
$$P(A_1|B) = \frac{(1/3) \cdot 0}{1/2} = 0.$$

Similarly, $P(A_2|B) = 1/3$, $P(A_3|B) = 2/3$.

■ **Q:** Given that 1st coin is gold, what is the probability that 2nd coin is gold?

- ▣ Let $C = \{\text{2nd coin is gold}\}$.
$$P(B \cap C|A_k) = \begin{cases} 0, & \text{if } k = 1, \\ 0, & \text{if } k = 2, \\ 1, & \text{if } k = 3. \end{cases}$$

- ▣
$$P(B \cap C) = \frac{1}{3} \cdot 0 + \frac{1}{3} \cdot 0 + \frac{1}{3} \cdot 1 = \frac{1}{3}.$$

$$P(C|B) = \frac{P(B \cap C)}{P(B)} = \frac{1/3}{1/2} = \frac{2}{3}.$$

➤ **Q:** Why are the 3 formulas useful in calculating probabilities? (Note: They all benefit from conditional probabilities.)

Ans: (i) 繁 → 簡 & 簡 & ... & 簡;

(ii) 簡 = conditioning because the sample space is reduced from Ω to a smaller set. (For example, in many cases, $P(B|A_j)$'s are known or are easier to find)

• Odds and Conditional Odds

➤ The odds of an event A :

$$o(A) \equiv \frac{P(A)}{P(A^c)} = \frac{P(A)}{1 - P(A)}$$

➤ The odd of event B given A :

$$o(B|A) \equiv \frac{P(B|A)}{P(B^c|A)}$$

and

$$o(B|A) = o(B) \times \frac{P(A|B)}{P(A|B^c)}$$

❖ **Reading:** textbook, Sec 3.1, 3.2, 3.3, 3.5

Independence

• **Definition (independence for 2-events case):** Two events A and B are said to be *independent* if and only if

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

Otherwise, they are said to be *dependent*.

➤ **Notes.** For independent events A and B , if $P(A) > 0$, then

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

similarly, if $P(B) > 0$, $P(A|B) = P(A)$.

Q: How to interpret the equality of the conditional and unconditional probabilities?

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➤ **Example (Sampling 2 balls, LNp.3-6)** A and B were independent for sampling with replacement, but dependent for sampling without replacement.

➤ **Example (Cards):** If a card is selected from a standard deck, let

▪ $A = \{\text{ace}\}$ and $B = \{\text{spade}\}$. Then,

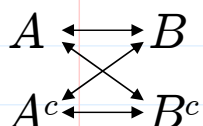
$$P(A) = \frac{4}{52} = \frac{1}{13}, \quad P(B) = \frac{13}{52} = \frac{1}{4},$$

$$P(A \cap B) = \frac{1}{52} = P(A)P(B)$$

\Rightarrow Face and Suit are independent

➤ **Theorem (Independence and Complements, 2-events case).**

If A and B are independent, then so are A^c and B .

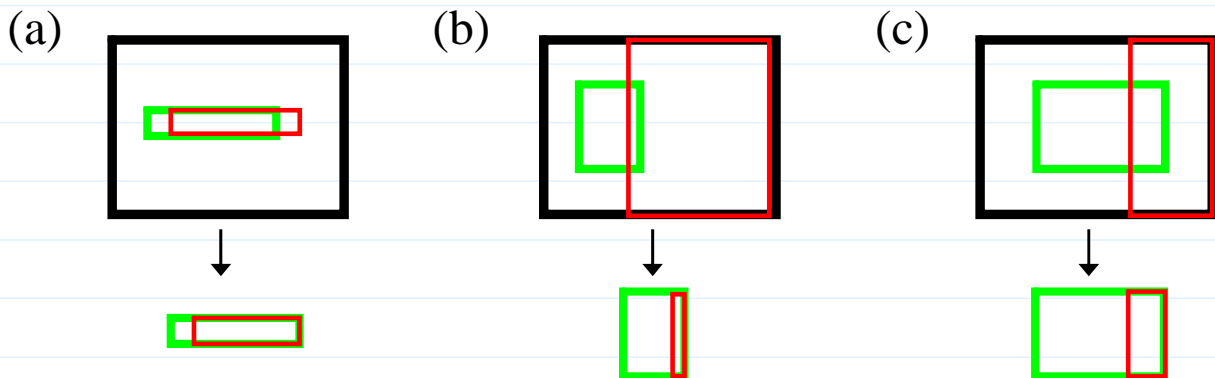


▪ **Corollary:** If A and B^c are independent, as are A^c and B^c

- Corellary: If A and B are independent and $0 < P(A) < 1$, $0 < P(B) < 1$, then
- $$P(B) = P(B|A) = P(B|A^c),$$
- $$P(B^c) = P(B^c|A) = P(B^c|A^c),$$
- $$P(A) = P(A|B) = P(A|B^c),$$
- $$P(A^c) = P(A^c|B) = P(A^c|B^c).$$

Q: What do these equalities say?

➤ **Q:** Which of the following graphs represents “green and red events are independent”?



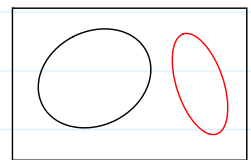
Q: Let green event = {graduate from Tsing-Hua University},
red event = {your future dream will come true}.

Which of the graphs would you prefer?

➤ Theorem (Independence and Mutually Exclusive).

If A and B are mutually exclusive and $P(A) > 0$, $P(B) > 0$, then A and B are dependent since

$$P(B|A) = 0 \neq P(B).$$



- Definition (independence for n -events case). Events A_1, \dots, A_n are said to be *pairwise independent* iff

$$P(A_i \cap A_j) = P(A_i)P(A_j),$$

for all $1 \leq i < j \leq n$; A_1, \dots, A_n are said to be *mutually independent* iff

$$P(A_i \cap A_j) = P(A_i)P(A_j), \quad \text{for } 1 \leq i < j \leq n,$$

$$P(A_i \cap A_j \cap A_k) = P(A_i)P(A_j)P(A_k), \quad \text{for } 1 \leq i < j < k \leq n,$$

...

$$P(A_{i_1} \cap \dots \cap A_{i_k}) = P(A_{i_1}) \dots P(A_{i_k}), \quad \text{for } 1 \leq i_1 < \dots < i_k \leq n,$$

where $k=2, \dots, n$.

➤ Note:

- Mutual independence implies pairwise independence; but, the converse statement is usually not true.
- “ n events are independent” means “mutually independent”

➤ Example (Sampling With Replacement)

- A sample of n balls is drawn with replacement from an urn containing R red and $N-R$ white balls
- Let $A_k = \{\text{red on the } k^{\text{th}} \text{ draw}\}$, then $P(A_k) = R/N$, $k=1, \dots, n$.
- For all $1 \leq i_1 < \dots < i_k \leq n$, where $k=2, \dots, n$,

$$P(A_{i_1} \cap \dots \cap A_{i_k}) = \frac{R^k N^{n-k}}{N^n} = \left(\frac{R}{N}\right)^k = P(A_{i_1}) \cdots P(A_{i_k}),$$

$\Rightarrow A_1, \dots, A_n$ are mutually independent

➤ Example. Draw one card from a standard deck.

- Let $A = \{\text{Spades or Clubs}\}$,
 $B = \{\text{Hearts or Clubs}\}$,
 $C = \{\text{Diamonds or Clubs}\}$.
- $P(A) = 26/52 = 1/2$, similarly, $P(B) = P(C) = 1/2$.
- $P(A \cap B) = P(\{\text{Clubs}\}) = \frac{13}{52} = \frac{1}{4} = P(A)P(B)$, similarly,
 $P(A \cap C) = 1/4 = P(A)P(C)$, $P(B \cap C) = 1/4 = P(B)P(C)$.
 $\Rightarrow A, B$, and C are pairwise independent

- However,
 $P(A \cap B \cap C) = P(\{\text{Clubs}\}) = \frac{1}{4} \neq \frac{1}{8} = P(A)P(B)P(C)$,
 $\Rightarrow A, B$, and C are *not* mutually independent

➤ Theorem (Independence and Complements, n -events case).

A_1, \dots, A_n are mutually independent if and only if

$$P(B_1 \cap \dots \cap B_n) = P(B_1) \cdots P(B_n),$$

where B_i is either A_i or A_i^c , for $i=1, \dots, n$.

- Example (Series and Parallel Connections of Relays).

Series Connection $S \rightarrow \text{---} R_1 \text{---} R_2 \text{---} R_3 \rightarrow E$

Parallel Connection $S \rightarrow \left[\begin{array}{c} \text{---} R_1 \text{---} \\ \text{---} R_2 \text{---} \\ \text{---} R_3 \text{---} \end{array} \right] \rightarrow E$

▣ The Story. For n electrical relays R_1, \dots, R_n , let

$$A_k = \{R_k \text{ works properly}\},$$

$k=1, \dots, n$, and suppose that A_1, \dots, A_n are independent.

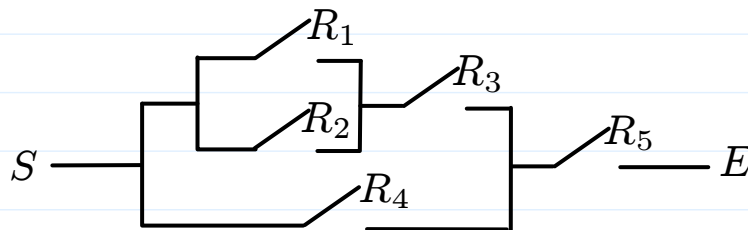
▣ Series Connection. The probability that current can flow from S to E (which corresponds to the event $A_1 \cap \dots \cap A_n$) is

$$P(A_1 \cap \dots \cap A_n) = P(A_1) \cdots P(A_n).$$

▣ Parallel Connection. The probability that current can flow from S to E (which corresponds to the event $A_1 \cup \dots \cup A_n$) is

$$\begin{aligned} P(A_1 \cup \dots \cup A_n) &= 1 - P(A_1^c \cap \dots \cap A_n^c) \\ &= 1 - P(A_1^c) \cdots P(A_n^c) = 1 - \prod_{k=1}^n [1 - P(A_k)] \end{aligned}$$

▣ Combination of Series and Parallel Connections



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➤ Theorem. If A_1, \dots, A_n are mutually independent and B_1, \dots, B_m , $m \leq n$, are formed by taking unions or intersections of mutually exclusive subgroups of A_1, \dots, A_n , then B_1, \dots, B_m are independent.

• Definition (conditional independence): Events B_1, \dots, B_n are (pairwise or mutually) independent under the probability measure $P(\cdot|A)$

➤ e.g., B_1 and B_2 are conditionally independent given A iff

$$P(B_1 \cap B_2 | A) = P(B_1 | A)P(B_2 | A),$$

or, equivalently,

$$P(B_1 | B_2 \cap A) = P(B_1 | A).$$

➤ Example (Gold Coins):

■ The Story.

- ▣ Box i contains i gold coins and $k-i$ silver coins, $i=0,1,\dots,k$.
- ▣ Experiment: (i) Select a box at random and, (ii) draw coins *with replacement* from the box

■ **Q:** Given that first n draws are all gold, what is the probability that $(n+1)^{\text{st}}$ draw is gold?

- ▣ Let $A_i = \{\text{Box } i \text{ is selected}\}$, $B = \{\text{first } n \text{ draws are gold}\}$,
 $C = \{(n+1)^{\text{st}} \text{ draw is gold}\}$
- ▣ By law of total probability,

$$P(C|B) = \sum_{i=0}^k P(A_i|B)P(C|A_i \cap B)$$

- ▣ Because B and C are conditionally independent given A_i ,

$$P(C|A_i \cap B) = P(C|A_i) = i/k$$

- ▣ By Bayes' rule,

$$P(A_i|B) = \frac{P(A_i)P(B|A_i)}{\sum_{j=0}^k P(A_j)P(B|A_j)} = \frac{[1/(k+1)](i/k)^n}{\sum_{j=0}^k [1/(k+1)](j/k)^n}$$

- ▣ Hence, $P(C|B) = \frac{\sum_{i=0}^k (i/k)^{n+1}}{\sum_{i=0}^k (i/k)^n}$

■ **Q:** Are the two events B and C independent?