

Intuition for Stats

CLT Red Flags in Word Problems

1. Sampling from a population repeatedly

Look for phrases like:

- "A sample of size n is taken..."
- "The average of 50 measurements..."
- "Suppose 100 items are selected at random..."

These suggest you're working with **sample statistics**, which is where the CLT comes in.

2. You're asked about a probability involving a sample mean

Key phrases:

- "What is the probability that the *average* weight is more than 150 lbs?"
- "Find the probability that the *mean* delivery time is less than 3 days."
- "Determine the probability that the *sample mean* falls between..."

This is a clear sign you need to apply CLT to the **distribution of sample means**.

3. The population distribution is NOT normal or unknown

CLT becomes relevant **especially** when the population is not normally distributed (or you aren't told). The CLT tells you the **sampling distribution of the sample mean** will be approximately normal if the sample size is large.

4. Sample size n is "large" (typically $n \geq 30$)

Even if the population isn't normal, CLT allows you to assume normality for the sample mean when n is large. If it's mentioned that:

- "The sample size is 40."
- "100 observations were taken."

...CLT is likely in play.

5. You're given population mean and standard deviation

Phrases like:

- "The population has mean $\mu=100$ and standard deviation $\sigma=15$ "
- "The population standard deviation is known..."

Mean + standard deviation + sample size is a classic setup for applying:

$$Z = \frac{\bar{X} - \mu}{\sigma / \sqrt{n}}$$

Which is the **CLT Z-score formula for sample means**.

CLT Usually Not Involved When:

- You're dealing with a **single observation** from the population.
- You're finding probabilities about **counts**, not **averages**.
- You're working with exact small samples from small populations (may require **exact binomial**, **hypergeometric**, or **t-distribution**).

TL;DR: Use CLT when...

| Clue | Meaning |
|---|--------------------------------|
| "Sample mean" or "average" | CLT is likely |
| Large sample size $n \geq 30$ | Approx. normal by CLT |
| Population distribution not normal or not mentioned | CLT helps you assume normality |
| You're given μ , σ , and n , and asked about probability | Classic CLT setup |

How to Recognize When to Use Conditional Probability (Plain)

Conditional Probability Formula:

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

✓ Clues that *plain conditional probability* is needed:

1. Problem talks about "given that..."

- Look for wording like:

"What is the probability that the person is a smoker
given that they are over 50 years old?"

2. You know/are given $P(A \text{ and } B)$ and $P(B)$

- The problem provides both probabilities directly, and you just need to use the formula above.

3. No "updating" beliefs after new information

- You're not "reversing" known probabilities (like Bayes); you're directly finding how two events relate.

⊘ Not conditional probability if:

- You're trying to **update** probabilities based on new evidence (that's Bayes').
- No event is "given" in the wording.

🔍 How to Recognize When to Use Bayes' Theorem

Bayes' Theorem Formula:

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

✓ Clues that Bayes' Theorem is needed:

1. You're given "reverse" information

- The problem tells you about $P(B|A)$ (e.g., test positive if sick), but asks for $P(A|B)$ (e.g., sick if tested positive).
- Common phrasing:

"Given that the test was positive, what is the probability that the person has the disease?"

2. **You have prior ("before") probabilities and you must "update" after new evidence**

- For example:
 - Prior: 2% of people have a disease.
 - Evidence: Positive test.
 - Question: Now, what's the probability they *actually* have the disease?

3. **Problem mentions "false positives", "true positives", or "testing"**

- Testing problems (medical tests, lie detectors, defect tests) very often use Bayes.

4. **Multiple possible causes or sources**

- Problem says:

"There are 3 machines, each produces different % of defects. Given that an item is defective, what is the probability it came from machine 1?"

- → **Bayes' Theorem needed.**

Simple Comparison

| If you are asked for... | Use |
|---|-------------------------|
| Probability of A given B, and you know joint and marginal probabilities | Conditional Probability |
| Update your belief about A after observing B, especially in diagnosis, testing, or backward reasoning | Bayes' Theorem |

Quick Visual Memory Trick

- **Direct "given that"** → **Conditional probability** (simple formula).
- **Reverse thinking or updating beliefs** → **Bayes' theorem** (more involved).

Probability questions involving **marbles or balls** typically fall under **classical probability** where outcomes are **finite, countable, and equally likely**. Here's a breakdown of the logic you can follow:

General Steps for Marble/Ball Probability Problems:

1. Understand the Setup:

- How many marbles/balls total?
- What are the colors/types?
- Are the draws with or without replacement?

2. Define the Events:

- What event's probability are you trying to compute? (e.g., both red, at least one blue, etc.)

3. Count Favorable Outcomes:

- How many ways can the desired event occur?

4. Count Total Outcomes:

- Total number of possible outcomes (depends on replacement and number of draws).

5. Apply the Basic Formula:

$$\text{Probability} = \frac{\text{Number of favorable outcomes}}{\text{Total number of possible outcomes}}$$

Key Variations and What to Look For:

| Type of Problem | Clue Words | What to Do |
|--------------------------------|---------------------------------|---|
| Without Replacement | "not replaced", "removed" | The total and probabilities change after each draw. |
| With Replacement | "replaced", "put back" | Probabilities stay the same for each draw. |
| Combinations | "choose 2", "randomly select k" | Use $\binom{n}{k}$ for selecting without order. |
| Conditional Probability | "given that", "if known that" | Use conditional probability formula: $P(A B) = P(A \cap B) / P(B)$ |

| | | |
|-------------------------------------|---------------------------------|---|
| Expected Value (advanced) | "on average", "expected number" | Multiply each outcome by its probability and sum. |
|-------------------------------------|---------------------------------|---|

Example Logic (Simple):

Q: A bag has 5 red and 3 blue balls. Two balls are drawn *without replacement*. What's the probability both are red?

- Total balls = 8
- Drawing 2 balls without replacement:

$$P(\text{both red}) = \frac{5}{8} \cdot \frac{4}{7} = \frac{20}{56} = \frac{5}{14}$$

1. Bernoulli Random Variable

Definition:

Single trial with only **two outcomes**: success (1) or failure (0).

Recognize when:

- You are doing **just one trial**.
- Only two outcomes ("success/failure", "yes/no", "win/lose").

Key words:

- "One attempt"
- "Did the customer buy the product: yes or no?"
- "Did the patient survive: yes or no?"

Important:

- If it's only **one** trial → **Bernoulli**.

2. Binomial Distribution

Definition:

Repeated **n** independent **Bernoulli trials**, counting the number of successes.

Recognize when:

- There are **multiple independent trials** (fixed number n).
- Each trial has **only two outcomes**.
- **Same probability** of success each time.

Key words:

- "Fixed number of trials"
- "5 customers come in. What is the probability 3 buy something?"
- "In 10 shots, probability of scoring exactly 7 goals."

Important:

- If there's n **independent, identical trials** → **Binomial**.
- Formula involves combinations:

$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$$

3. Poisson Distribution

Definition:

Counts the number of events happening in a **fixed interval** (time, area, volume), where events occur **independently** at a constant average rate.

Recognize when:

- **No fixed number of trials** — events just happen randomly.
- Describing events **over time or space**.
- Events are rare.

Key words:

- "per hour", "per day", "per km", "per page"
- "On average, 3 accidents occur per week."
- "How many calls are received at a call center in an hour?"

Important:

- **Poisson** is for **event counts** in **continuous domain** (time, space).

- Formula:

$$P(X = k) = \frac{e^{-\lambda} \lambda^k}{k!}$$

where λ = average number of events.

4. 🎯 Uniform Distribution

(Discrete Uniform)

Definition:

Each of a **finite number** of outcomes is **equally likely**.

Recognize when:

- All outcomes have **equal probability**.
- Picking at random from a set.

Key words:

- "Roll a die" (faces 1–6 are equally likely)
 - "Choose a card at random"
-

(Continuous Uniform)

Definition:

All real numbers between two values are **equally likely**.

Recognize when:

- Picking a **real number** at random between two limits.
- Continuous outcomes (not just countable).

Key words:

- "Arrival time is uniformly distributed between 2 PM and 4 PM."
- "Temperature is uniformly distributed between 20°C and 30°C."

Important:

- For continuous:

$$f(x) = \frac{1}{b-a}, \quad a \leq x \leq b$$

5. 🎯 Normal Distribution (Gaussian)

Definition:

A bell-shaped, symmetric distribution used for many natural phenomena.

Center = mean, **spread** = standard deviation.

Recognize when:

- The data is **continuous**.
- The problem talks about **averages, measurements, heights, weights, test scores**, etc.
- Symmetric around a center value.

Key words:

- "Normally distributed"
- "The height of students is normally distributed with mean 170 cm and standard deviation 10 cm."
- "Assume normal distribution."

Important:

- Z-scores are often used:

$$Z = \frac{X - \mu}{\sigma}$$

🎯 Bonus Tip

If a problem says "mean = μ , variance = σ^2 " and **large number of trials or samples** → **CLT (Central Limit Theorem)** is likely involved, even if the original distribution is not normal!