

Sampling Techniques: Comprehensive Viva Guide

Quick Reference for Oral Examination

1 Definitions and Core Concepts

Random Sampling

Definition: A sampling method where each element in the population has an *equal and independent* probability of being selected.

Key Properties:

- Probability of selection: $P(x_i) = \frac{1}{N}$ for population size N
- Each selection is independent of others
- Simple Random Sampling (SRS) is the foundation of statistical inference

Stratified Sampling

Definition: The population is divided into homogeneous subgroups (strata) based on a characteristic, and random samples are drawn from each stratum.

Key Properties:

- Population = $\bigcup_{h=1}^L S_h$ where S_h are disjoint strata
- Sample from each stratum: n_h from stratum h
- Total sample size: $n = \sum_{h=1}^L n_h$
- Ensures representation from all subgroups

Hybrid Sampling (CRITICAL)

Definition: A two-stage approach combining stratified sampling with random sampling within strata. First stratify, then apply different sampling rates or methods per stratum.

Key Characteristics:

- Combines advantages of both stratification and random sampling
- Allocation can be proportional or optimal
- Reduces variance compared to simple random sampling
- Flexible: can oversample rare strata

Stratified Clustering

Definition: A multi-stage technique where the population is first divided into strata, then clusters are formed within strata, and samples are drawn from selected clusters.

Structure:

- Stage 1: Divide population into strata
- Stage 2: Identify clusters within each stratum
- Stage 3: Sample clusters, then sample elements within clusters
- Useful for geographically dispersed populations

2 Mathematical Foundations

2.1 Random Sampling

Sample Mean Estimator:

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

Variance of Sample Mean:

$$\text{Var}(\bar{x}) = \frac{\sigma^2}{n} \left(1 - \frac{n}{N}\right) = \frac{\sigma^2}{n} \cdot \text{FPC}$$

where FPC = finite population correction = $(1 - \frac{n}{N})$

Standard Error:

$$SE(\bar{x}) = \sqrt{\frac{s^2}{n} \left(1 - \frac{n}{N}\right)}$$

2.2 Stratified Sampling

Stratified Mean Estimator:

$$\bar{x}_{st} = \sum_{h=1}^L W_h \bar{x}_h$$

where $W_h = \frac{N_h}{N}$ is the weight of stratum h , and \bar{x}_h is the sample mean in stratum h .

Variance of Stratified Mean:

$$\text{Var}(\bar{x}_{st}) = \sum_{h=1}^L W_h^2 \frac{\sigma_h^2}{n_h} \left(1 - \frac{n_h}{N_h}\right)$$

Allocation Methods:

- *Proportional:* $n_h = n \cdot W_h = n \cdot \frac{N_h}{N}$
- *Optimal (Neyman):* $n_h = n \cdot \frac{N_h \sigma_h}{\sum_{k=1}^L N_k \sigma_k}$
- Optimal allocation minimizes variance for fixed cost

2.3 Hybrid Sampling

Hybrid sampling variance depends on the specific combination used. For stratified random sampling hybrid:

$$\text{Var}(\bar{x}_{hybrid}) \leq \text{Var}(\bar{x}_{SRS})$$

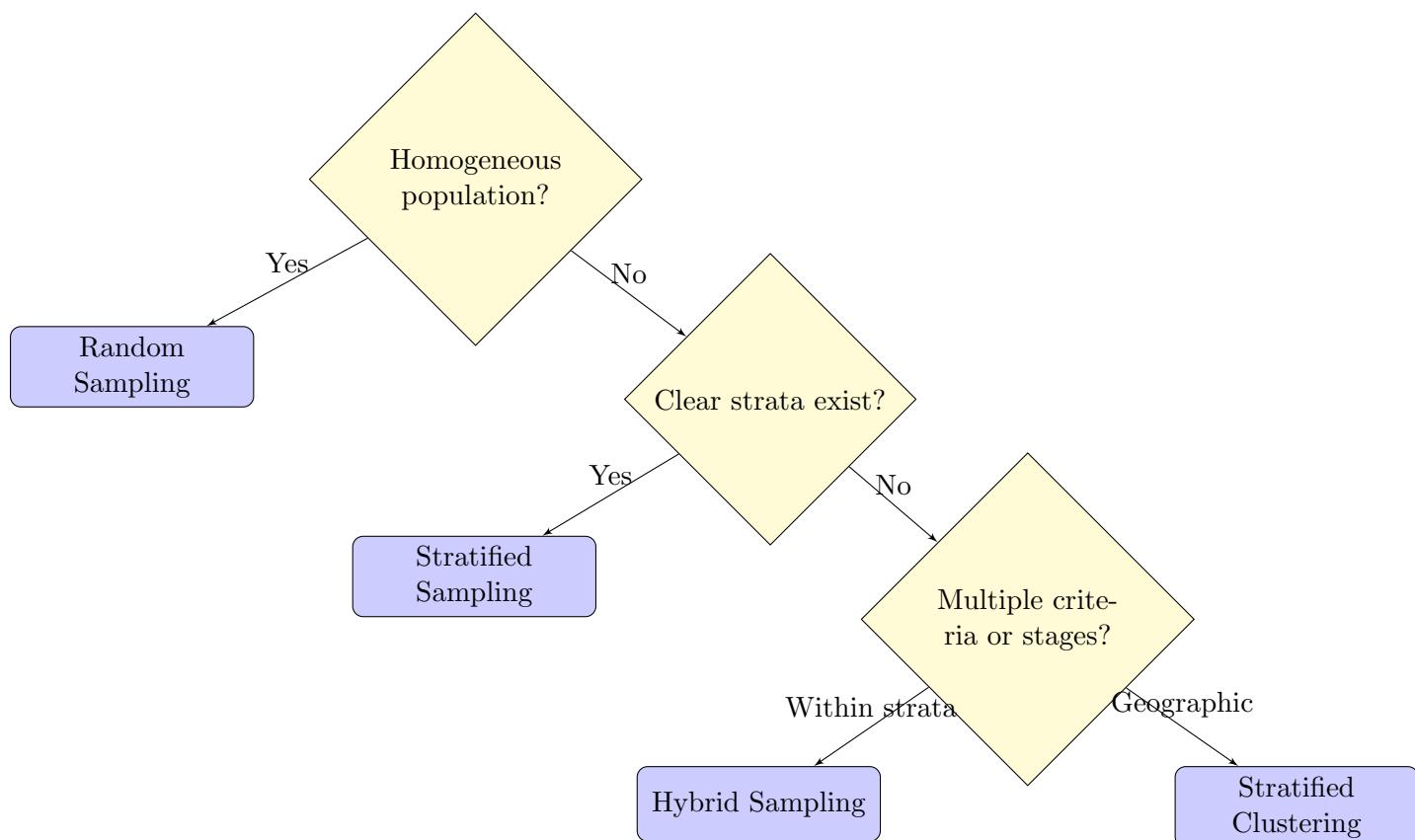
The reduction factor depends on stratum homogeneity and allocation efficiency.

3 Practical Understanding

3.1 When to Use Each Technique

Technique	Best Used When	Example
Random Sampling	Homogeneous population, no prior knowledge	Selecting students from a single class
Stratified	Known heterogeneous subgroups, want precision	Income survey across age groups
Hybrid	Need flexibility, multiple characteristics	Customer satisfaction with regional and age stratification
Stratified Clustering	Large geographic area, cost constraints	National health survey by state then city

3.2 Decision Flowchart



4 Common Viva Questions with Detailed Answers

Q1: What is the main advantage of stratified sampling over random sampling?

Answer: Stratified sampling provides *greater precision* (lower variance) than simple random sampling for the same sample size. This occurs because:

- Within-stratum variance is smaller than population variance
- We ensure representation from all subgroups
- Mathematical proof: $\text{Var}(\bar{x}_{st}) \leq \text{Var}(\bar{x}_{SRS})$ when strata are homogeneous
- The gain is maximum when strata are internally homogeneous but differ from each other

Q2: Explain hybrid sampling and why it's useful.

Answer: Hybrid sampling combines multiple sampling techniques, typically stratified sampling followed by different methods within strata.

Key benefits:

- *Flexibility:* Can use different sampling rates for different strata
- *Efficiency:* Oversample rare but important subgroups
- *Cost-effective:* Apply expensive methods only where needed
- *Precision:* Maintains stratification benefits while adapting to constraints

Example: In a national survey, stratify by region, then use cluster sampling in rural areas (cost-effective) and simple random sampling in urban areas (more precise).

Q3: How does stratified clustering differ from stratified sampling?

Answer:

- **Stratified Sampling:** Sample individuals directly from each stratum
- **Stratified Clustering:** Sample clusters from each stratum, then sample individuals from selected clusters

Key differences:

1. *Stages:* Stratified clustering is multi-stage; stratified sampling is single-stage
2. *Sampling unit:* Clusters first, then individuals vs. individuals directly
3. *Cost:* Clustering reduces travel/administrative costs
4. *Variance:* Clustering typically increases variance due to intra-cluster correlation

Q4: What is the finite population correction (FPC) and when is it important?

Answer: FPC = $(1 - \frac{n}{N})$ accounts for sampling without replacement from finite populations.

When important:

- When sampling fraction $\frac{n}{N} > 0.05$ (5% rule)
- Reduces estimated variance
- As $n \rightarrow N$, variance $\rightarrow 0$ (census)
- Often ignored for large populations where $\frac{n}{N} \approx 0$

Q5: Derive or explain the stratified mean estimator.

Answer: The stratified mean is a weighted average of stratum means:

$$\bar{x}_{st} = \sum_{h=1}^L W_h \bar{x}_h$$

Intuition:

- Each stratum contributes proportionally to its size in the population
- $W_h = \frac{N_h}{N}$ ensures proper weighting
- If we sampled proportionally, $\bar{x}_{st} = \bar{x}$ (equal to simple mean)
- Unbiased: $E(\bar{x}_{st}) = \mu$ (population mean)

Q6: What is optimal (Neyman) allocation and when should you use it?

Answer: Neyman allocation minimizes variance for fixed total sample size:

$$n_h = n \cdot \frac{N_h \sigma_h}{\sum_{k=1}^L N_k \sigma_k}$$

Interpretation:

- Allocate more samples to larger strata (N_h)
- Allocate more samples to more variable strata (σ_h)
- *Use when:* You know or can estimate stratum variances and want minimum variance
- *Challenge:* Requires prior knowledge of σ_h

Q7: Why might stratified clustering be preferred over stratified sampling despite higher variance?

Answer: Cost and logistics:

- Dramatically reduces travel and data collection costs
- More practical for geographically dispersed populations
- Easier administrative implementation
- Can sample more elements for the same budget

Trade-off: Accept higher variance per unit in exchange for larger affordable sample size, potentially achieving lower overall variance.

Q8: How do you determine the number of strata in stratified sampling?

Answer: Consider:

1. **Heterogeneity:** More strata if population is highly variable
2. **Sample size:** Need sufficient $n_h \geq 2$ in each stratum for variance estimation
3. **Known characteristics:** Use natural divisions (age groups, regions)
4. **Diminishing returns:** Beyond 6-8 strata, gains are often minimal
5. **Practical constraints:** Cost and complexity of managing many strata

Rule of thumb: More homogeneous within strata than between strata.

Q9: Compare proportional vs. optimal allocation.

Answer:

Proportional ($n_h \propto N_h$):

- Simpler to implement
- Self-weighting: $\bar{x}_{st} = \bar{x}$
- Good when stratum variances are similar
- No prior variance knowledge needed

Optimal ($n_h \propto N_h \sigma_h$):

- Minimizes variance
- Better when stratum variances differ substantially
- Requires variance estimates
- More complex weighting in estimation

Q10: What are the assumptions for valid stratified sampling?**Answer:**

1. **Disjoint strata:** Each population element belongs to exactly one stratum
2. **Exhaustive:** Strata cover entire population
3. **Known stratum sizes:** N_h must be known for all h
4. **Random sampling within strata:** Each element in a stratum has equal probability
5. **Independent selections:** Selections across strata are independent

5 Comparison Points

5.1 Variance Hierarchy (Best to Worst)

For well-designed surveys with homogeneous strata:

$$\text{Var}(\text{Optimal Stratified}) < \text{Var}(\text{Proportional Stratified}) < \text{Var}(\text{Random}) < \text{Var}(\text{Clustering})$$

5.2 Cost Hierarchy (Cheapest to Most Expensive)

$$\text{Clustering} < \text{Stratified Clustering} < \text{Stratified} < \text{Random}$$

Key insight: Trade-off between statistical efficiency and practical cost.

6 Real-World Applications

1. **Political Polling:** Stratify by state/region, age, gender; hybrid approach with different methods for urban/rural
2. **Quality Control:** Stratify by production line, shift, or product type
3. **Medical Research:** Stratify by age groups, severity of condition; ensures adequate representation
4. **Market Research:** Stratified clustering by geographic region then retail locations
5. **Educational Assessment:** Stratify by school type, grade level; random sampling within strata
6. **Agricultural Surveys:** Stratified clustering by county then farms; accounts for regional variation

7 Common Pitfalls and Misconceptions

Common Mistakes

1. **Confusing stratified and cluster sampling:**
 - Stratified: Sample from *all* strata
 - Cluster: Sample *some* clusters, then all or sample within
2. **Forgetting weights in stratified estimation:** Must use W_h ; simple averaging gives wrong answer if allocation isn't proportional
3. **Ignoring FPC:** Can overestimate variance when sampling fraction is substantial
4. **Using too many strata:** Results in small n_h , unstable variance estimates
5. **Poor stratification criteria:** Choosing variables unrelated to outcome reduces efficiency gains
6. **Assuming clustering reduces variance:** Clustering typically *increases* variance due to within-cluster homogeneity
7. **Not checking stratum assumptions:** Ensure strata are truly disjoint and exhaustive
8. **Optimal allocation without variance knowledge:** Can't use Neyman allocation without estimating σ_h

Key Takeaways

- **Stratification reduces variance** when strata are homogeneous internally
- **Hybrid sampling offers flexibility** to combine methods optimally
- **Clustering trades precision for cost** in geographically dispersed surveys
- **Stratified clustering combines both approaches** for practical large-scale surveys
- **Choice depends on:** population structure, budget, precision requirements