

DSA 507: Statistics for Data Science and AI

Answer all questions. Each sub-question carries equal marks. Use clear reasoning and, where relevant, refer to the lecture content: <https://usjstat.github.io/courses/dsa507/>

1. Structured vs Unstructured Data

- (a) (2 points) Define structured data and provide an example.

Solution: Structured data refers to data that is organized in a fixed format such as rows and columns, often stored in tables or data frames.
Example: The 'cricket_data' data frame with players, runs, strike rate, and captaincy status.

- (b) (2 points) List two main types of structured data along with one example for each.

Solution: - Numeric: e.g., Number of runs scored (continuous or discrete).
- Categorical: e.g., Player role such as "Bowler", "Batsman" (nominal), or education level (ordinal).

- (c) (3 points) Explain the significance of converting unstructured data into a structured form for statistical analysis.

Solution: Converting unstructured data into structured form enables the application of statistical methods such as summary statistics, modeling, and visualization. Structured formats are necessary for tools like R or Python to perform EDA.

2. Estimates of Location and Variability

- (a) (2 points) Explain the difference between mean, median, and trimmed mean using your own words.

Solution: - Mean: Arithmetic average of all values.
- Median: Middle value when data is sorted.
- Trimmed Mean: Mean calculated after removing extreme values from both ends.

- (b) (2 points) Why is the median considered a robust estimator?

Solution: The median is robust because it is not affected by outliers or extreme values. It only depends on the order of data.

- (c) (2 points) Explain what the interquartile range (IQR) measures.

Solution: IQR measures the range of the middle 50% of the data. It is the difference between the 75th and 25th percentiles.

- (d) (2 points) Which of the following is most affected by an outlier: mean, median, mode?

Solution: Correct answer: Mean

3. Visualizing Distributions

- (a) (3 points) Describe what each of the following visualizations shows: histogram, density plot, boxplot.

Solution: - Histogram: Shows frequency distribution across bins.
- Density Plot: Smoothed curve showing distribution shape.
- Boxplot: Summarizes median, quartiles, and highlights outliers.

- (b) (3 points) In a boxplot, what do the "whiskers" represent?

Solution: The whiskers extend to the smallest and largest values within $1.5 * \text{IQR}$ from the lower and upper quartiles.

- (c) (2 points) Explain how outliers are identified in a boxplot.

Solution: Outliers are identified as data points that fall outside the whiskers (i.e., beyond $1.5 * \text{IQR}$ from Q1 or Q3).

4. Exploring Categorical Data

- (a) (2 points) Differentiate between nominal, ordinal, and binary categorical data with one example each.

Solution: - Nominal: Categories with no order (e.g., team name).
- Ordinal: Categories with a clear order (e.g., education level).
- Binary: Two categories (e.g., captaincy status: Yes/No).

- (b) (2 points) Define expected value in the context of binary outcomes.

Solution: Expected value is the long-run average outcome. For binary variables (e.g., Win=1, Loss=0), it's the proportion of Wins.

- (c) (2 points) Explain why bar plots are preferred over pie charts in professional reporting.

Solution: Bar plots are easier to interpret, more accurate for comparison, and suitable for publications. Pie charts are harder to read and interpret.

5. Correlation and Multivariate Exploration

- (a) (2 points) What does a correlation coefficient of 0.95 imply about the relationship between two variables?

Solution: A correlation coefficient of 0.95 indicates a very strong positive linear relationship — as one variable increases, the other also increases.

- (b) (2 points) Match each visualization with its primary use:

- | | |
|-----------------|---|
| A. Violin Plot | 1. Handling overplotting for dense scatterplots |
| B. Contour Plot | 2. Distribution shape + central tendency |
| C. Hexbin Plot | 3. Topographical density view |

Your answers: A → ____, B → ____, C → ____

Solution: Correct matching:

A → 2,
B → 3,
C → 1

6. Population, Sample, and Sampling Methods

- (a) (2 points) Define population and sample using a sports example.

Solution: Population: All matches played in a football league season.
Sample: 50 randomly selected matches from that season.

- (b) (2 points) Explain the importance of random sampling in data science.

Solution: Random sampling ensures each unit has an equal chance of being selected, reducing bias and improving representativeness.

- (c) (3 points) Describe the difference between sampling with and without replacement.

Solution: With replacement: a unit can be selected multiple times.
Without replacement: once selected, a unit cannot be selected again.

- (d) (3 points) What is sample bias? Give one historical example (mentioned in the notes).

Solution: Sample bias occurs when the sample misrepresents the population. Example: Literary Digest's 1936 election poll using car and phone owners.

7. Selection Bias and Regression to the Mean

- (a) (2 points) Define selection bias and give one way to avoid it.

Solution: Selection bias arises when the sample does not represent the population due to how it's selected.
Avoid it by using random selection and holdout validation.

- (b) (2 points) Explain the concept of regression to the mean.

Solution: Extreme observations tend to be followed by more average ones due to random variation.

8. Sampling Distributions and Central Limit Theorem (CLT)

- (a) (2 points) What is a sampling distribution?

Solution: The distribution of a statistic (e.g., mean) over many repeated samples from a population.

- (b) (2 points) What happens to the shape of the sampling distribution of the mean as sample size increases?

Solution: It becomes more bell-shaped and less variable — closer to a normal distribution.

- (c) (3 points) State the Central Limit Theorem.

Solution: The sampling distribution of the sample mean approaches a normal distribution as sample size increases, regardless of the population's shape (provided it's not too skewed).

- (d) (3 points) Define standard error and explain its relation to sample size.

Solution: Standard error is the standard deviation of the sampling distribution. It decreases as sample size increases.

9. Bootstrap and Confidence Intervals

- (a) (2 points) What is the bootstrap method used for?

Solution: Estimating the sampling distribution of a statistic by resampling with replacement from the observed data.

- (b) (2 points) Outline the steps of bootstrapping a mean.

Solution: (1) Resample data with replacement, (2) compute mean, (3) repeat many times, (4) analyze distribution of means.

- (c) (3 points) How do you construct a 90% bootstrap confidence interval for the mean?

Solution: Take 1000+ bootstrap samples, compute the mean for each, then take the 5th and 95th percentiles as interval bounds.

- (d) (3 points) What does a 90% confidence interval imply?

Solution: If we repeated the sampling process many times, about 90% of the constructed intervals would contain the true population mean.

10. Normal Distribution and Z-Scores

- (a) (2 points) State two properties of the normal distribution.

Solution: It is symmetric and bell-shaped; defined by mean μ and standard deviation σ .

- (b) (2 points) What is the empirical rule? List the percentages for $\pm 1\sigma$, $\pm 2\sigma$, and $\pm 3\sigma$.

Solution: 68% within $\pm 1\sigma$, 95% within $\pm 2\sigma$, 99.7% within $\pm 3\sigma$.

- (c) (2 points) Convert a height of 220 cm to a z-score given $\mu = 200$, $\sigma = 10$.

Solution: $z = \frac{220-200}{10} = 2$

- (d) (2 points) Does standardizing a dataset make it normally distributed? Briefly explain.

Solution: No. Standardization only rescales the data; it does not change the underlying distribution shape.

11. Assessing Normality and Central Limit Theorem in Practice

- (a) (2 points) What does a Q-Q plot show? How do you interpret it?

Solution: A Q-Q plot compares sample quantiles to theoretical normal quantiles. Points on a straight diagonal line suggest normality.

- (b) (2 points) How can a histogram of sample means provide evidence for the Central Limit Theorem?

Solution: Even if the original data is skewed, the histogram of sample means tends toward a normal shape as sample size increases.

- (c) (3 points) Why does the sampling distribution of the mean tend to be more bell-shaped than the original data distribution?

Solution: Because averaging reduces variability and extreme values, leading to a more symmetric and normal-like shape (CLT).

- (d) (3 points) Suppose a right-skewed dataset has a mean of 30 and SD of 10. What is the expected shape of the sampling distribution of the mean for samples of size 50?

Solution: Approximately normal-shaped due to the Central Limit Theorem.

12. Student's t-Distribution

- (a) (2 points) Write down the formula for the t-statistic.

Solution: $t = \frac{\bar{x} - \mu}{s/\sqrt{n}}$

- (b) (2 points) When should we use the t-distribution instead of the normal distribution?

Solution: When the sample size is small and population standard deviation is unknown.

- (c) (2 points) What happens to the t-distribution as degrees of freedom increase?

Solution: It approaches the standard normal distribution.

- (d) (2 points) Why does the t-distribution have heavier tails than the normal distribution?

Solution: To account for the additional uncertainty when estimating the standard deviation from small samples.

13. Discrete Probability Distributions

- (a) (2 points) Define the binomial distribution and provide its mean and variance.

Solution: Models number of successes in n trials with probability π . Mean = $n\pi$, Variance = $n\pi(1 - \pi)$

- (b) (2 points) What does the chi-square distribution test?

Solution: It tests whether observed counts significantly differ from expected counts under a null hypothesis.

- (c) (2 points) State the formula for the chi-square statistic.

Solution: $\chi^2 = \sum \frac{(O-E)^2}{E}$

- (d) (2 points) Define the F-distribution and state one common application.

Solution: The ratio of two variances; used in ANOVA to compare group means.

14. Modeling Events Over Time or Space

- (a) (2 points) When is the Poisson distribution appropriate?

Solution: When modeling the number of events in a fixed time or space interval, assuming constant rate and independence.

- (b) (2 points) What is the relationship between the Poisson and exponential distributions?

Solution: Poisson models count of events; exponential models time between those events.

- (c) (2 points) What parameter of the Weibull distribution allows for increasing or decreasing failure rates?

Solution: The shape parameter β .

- (d) (2 points) State two practical applications of the Weibull distribution.

Solution: Reliability engineering and failure analysis.

15. A/B Testing and Experimental Design

- (a) (2 points) Define A/B testing and explain its relevance in data science with an example.

Solution: A/B testing compares two treatments or strategies to determine which performs better. In sports analytics, it helps evaluate training programs, equipment, or tactics.

- (b) (2 points) Explain the importance of randomization in experimental design.

Solution: Randomization eliminates selection bias and ensures that differences in outcomes are due to the treatment and not pre-existing differences.

- (c) (2 points) Why is a control group necessary in a statistical experiment?

Solution: A control group serves as a baseline, allowing researchers to isolate the treatment effect by comparison.

- (d) (2 points) What is the difference between single-blind and double-blind experiments?

Solution: In single-blind, participants do not know which treatment they receive; in double-blind, both participants and researchers are unaware.

- (e) (2 points) Describe one ethical consideration when conducting experiments involving human subjects.

Solution: Informed consent is critical—participants must know and agree to the nature of the experiment, especially if it can affect their performance or well-being.

16. Hypothesis Testing and p-Values

- (a) (2 points) Define null hypothesis and alternative hypothesis with an example.

Solution: Null: No difference exists (e.g., Strategy A = Strategy B). Alternative: A difference exists (e.g., Strategy A \neq Strategy B).

- (b) (2 points) What is a p-value?

Solution: The probability of observing a result as extreme as the actual one, assuming the null hypothesis is true.

- (c) (2 points) What does it mean if a result is statistically significant at $\alpha = 0.05$?

Solution: It means the p-value is less than 0.05, suggesting the observed result is unlikely due to chance alone.

- (d) (2 points) Explain the difference between a one-tailed and two-tailed test.

Solution: A one-tailed test checks for an effect in one direction only; a two-tailed test checks for any difference regardless of direction.

- (e) (2 points) What is a Type I error? What is a Type II error?

Solution: Type I: Rejecting a true null hypothesis (false positive). Type II: Failing to reject a false null hypothesis (false negative).

17. Permutation Tests and Resampling

- (a) (2 points) Describe the purpose of a permutation test.

Solution: To test the null hypothesis by evaluating how likely the observed statistic is under all possible rearrangements of group labels.

- (b) (2 points) What is the difference between bootstrapping and permutation testing?

Solution: Bootstrapping estimates variability; permutation tests assess significance by reshuffling data.

- (c) (2 points) In what type of scenario is a permutation test especially useful?

Solution: When the distributional assumptions (e.g., normality) of parametric tests are not met.

- (d) (2 points) How is the permutation distribution generated?

Solution: By repeatedly shuffling the data labels, reassigning to groups, and computing the test statistic each time.

- (e) (2 points) What does it mean if the observed difference lies in the extreme tail of the permutation distribution?

Solution: It suggests the result is unlikely due to chance, and the treatment effect may be significant.

18. Multiple Testing and Adjustments

- (a) (2 points) What is alpha inflation, and why does it occur during multiple hypothesis testing?

Solution: Alpha inflation refers to the increased probability of at least one Type I error when performing multiple tests. It occurs because each test carries its own chance of a false positive.

- (b) (2 points) Calculate the probability of making at least one Type I error when performing 10 independent tests at $\alpha = 0.05$.

Solution: $1 - (1 - 0.05)^{10} \approx 0.4013$

- (c) (2 points) Explain how the Bonferroni adjustment modifies the significance level.

Solution: It divides the overall alpha by the number of tests, reducing the significance threshold for each individual test.

- (d) (2 points) What is the primary advantage of the False Discovery Rate (FDR) approach over Bonferroni in exploratory research?

Solution: FDR balances false positives with discovery, allowing more findings in exploratory contexts without overly penalizing multiple tests.

- (e) (2 points) Write the formula for Tukey's HSD and identify what each symbol means.

Solution: $HSD = q_{\alpha, k, df_{error}} \times \sqrt{\frac{MS_{error}}{n}}$, where q is the studentized range statistic, k number of groups, df_{error} error degrees of freedom, MS_{error} mean square error, and n observations per group.

19. One-Way ANOVA and the F-Statistic

- (a) (2 points) Define the decomposition of variance in the context of ANOVA.

Solution: It breaks an observation into grand mean, treatment effect, and residual: $Y_{ij} = \bar{Y}_{..} + \tau_j + \epsilon_{ij}$.

- (b) (2 points) Write the formula for the F-statistic in One-Way ANOVA.

Solution: $F = \frac{MSB}{MSW} = \frac{SSB/(k-1)}{SSW/(N-k)}$

- (c) (2 points) What does a large F value indicate in ANOVA?

Solution: That between-group variability is large relative to within-group variability, suggesting group means differ significantly.

- (d) (2 points) When do we reject the null hypothesis in ANOVA?

Solution: When the p-value associated with the F-statistic is less than the significance level α , often 0.05.

20. Post-Hoc Testing and Tukey's HSD

- (a) (2 points) What is the purpose of a post-hoc test in ANOVA?

Solution: To determine which specific group means differ after finding a significant ANOVA result.

- (b) (2 points) Write the formula for Tukey's HSD.

Solution: $HSD = q_{\alpha, k, df_{error}} \times \sqrt{\frac{MS_{error}}{n}}$

- (c) (2 points) When is Tukey's HSD preferred over pairwise t-tests?

Solution: When multiple comparisons are needed after ANOVA, to control the family-wise error rate.

- (d) (2 points) What does it mean if the confidence interval from Tukey's test includes zero?

Solution: The two groups being compared are not significantly different.

- (e) (2 points) Why do we adjust for multiple comparisons?

Solution: To reduce the risk of Type I errors due to multiple hypothesis tests.

21. Two-Way ANOVA and Interaction Effects

- (a) (2 points) State the full model equation for a Two-Way ANOVA with interaction.

Solution: $Y_{ijk} = \mu + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \epsilon_{ijk}$

- (b) (2 points) What does a significant interaction term in Two-Way ANOVA suggest?

Solution: That the effect of one factor depends on the level of the other factor.

- (c) (2 points) What are the null hypotheses tested in a Two-Way ANOVA?

Solution: $H_{0A} : \alpha_i = 0$, $H_{0B} : \beta_j = 0$, $H_{0AB} : (\alpha\beta)_{ij} = 0$

- (d) (2 points) Why is it important to check interaction before interpreting main effects?

Solution: Because a significant interaction invalidates interpretation of main effects independently.

22. ANOVA Assumptions and Diagnostics

- (a) (2 points) List the key assumptions of ANOVA.

Solution: Independence, normality of residuals, and homogeneity of variances.

- (b) (2 points) What is the purpose of the Q-Q plot in ANOVA diagnostics?

Solution: To assess whether residuals are approximately normally distributed.

- (c) (2 points) What does a Residuals vs Fitted plot help assess?

Solution: Homoscedasticity—whether residual variance is constant across fitted values.

- (d) (2 points) What can you conclude if the residual plot shows a funnel shape?

Solution: Heteroscedasticity—variance increases or decreases with fitted values, violating ANOVA assumptions.

- (e) (2 points) What action can be taken if ANOVA assumptions are violated?

Solution: Use transformations, robust methods, or non-parametric alternatives.

23. Chi-Square Test of Independence

- (a) (2 points) Define the chi-square statistic and state its formula.

Solution: $\chi^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}}$

- (b) (2 points) What are the degrees of freedom in a chi-square test for an $r \times c$ table?

Solution: $(r - 1)(c - 1)$

- (c) (2 points) In which situation should you use Fisher's Exact Test instead of a chi-square test?

Solution: When expected cell counts are very small (typically less than 5), especially in 2x2 tables.

- (d) (2 points) What does a large chi-square statistic suggest?

Solution: That observed counts differ significantly from expected counts, suggesting dependence between variables.

24. Expected Counts and Pearson Residuals

- (a) (2 points) Explain how expected counts are computed in a chi-square test.

Solution: $E_{ij} = \frac{(\text{row total})_i \times (\text{column total})_j}{\text{grand total}}$

- (b) (2 points) Define a Pearson residual and provide its formula.

Solution: $R_{ij} = \frac{O_{ij} - E_{ij}}{\sqrt{E_{ij}}}$

- (c) (2 points) What do large absolute values of Pearson residuals indicate?

Solution: Cells that contribute significantly to the chi-square statistic.

- (d) (2 points) Why is it important to check residuals even when p-value is significant?

Solution: To understand which specific combinations of categories deviate from the null model.

- (e) (2 points) Calculate the expected count for Player A and "Made" given 3 players with 100 attempts each, and total made = 240.

Solution: Expected = $\frac{100 \times 240}{300} = 80$

25. Chi-Square Test: Resampling vs. Theoretical

- (a) (2 points) What is the main idea of the permutation (resampling) test?

Solution: Randomly shuffle labels to generate a null distribution for the test statistic.

- (b) (2 points) What does the p-value represent in a permutation test?

Solution: Proportion of permuted test statistics that are at least as extreme as the observed statistic.

- (c) (2 points) Why might permutation p-values differ from theoretical p-values?

Solution: Permutation is data-driven and does not assume large-sample approximations.

- (d) (2 points) When are theoretical chi-square tests reliable?

Solution: When all expected cell counts are at least 5.

26. Multi-Arm Bandit(MAB) Algorithms

- (a) (2 points) What is the exploration vs. exploitation tradeoff in the context of the MAB problem?

Solution: Exploration involves trying different options to learn about their success rates; exploitation involves choosing the best-known option to maximize success.

- (b) (2 points) What role does the epsilon (ϵ) parameter play in the epsilon-greedy algorithm?

Solution: It controls the probability of exploring a random option. With probability ϵ , explore; with $1 - \epsilon$, exploit the best-known option.

- (c) (2 points) Define a ‘win’ in the context of the T20 batting pair MAB simulation.

Solution: A win is defined as the batting pair scoring at least 30 runs in the Powerplay (first 6 overs).

- (d) (2 points) In Thompson Sampling, how is the posterior of Beta distribution for each arm updated after each trial?

Solution: The α parameter of Beta prior is increased by 1 for a success; the β parameter of Beta prior is increased by 1 for a failure.

27. Power and Sample Size Analysis

- (a) (2 points) Define statistical power and its importance in hypothesis testing.

Solution: Power is the probability of correctly rejecting the null hypothesis when it is false. High power reduces the risk of Type II errors.

- (b) (2 points) Explain what happens to power when sample size increases, assuming fixed effect size and significance level.

Solution: Power increases as sample size increases, because the test becomes more sensitive to detecting the effect.

28. Simulation-Based Power Estimation

- (a) (2 points) Describe the basic steps of estimating power through simulation.

Solution: Generate samples from two groups, run hypothesis tests repeatedly, and compute the proportion of significant p-values.

- (b) (2 points) What is the advantage of using simulation-based methods over analytical methods for power analysis?

Solution: They allow estimation without strict distributional assumptions and can handle complex or real-world data scenarios.

- (c) (2 points) Explain why permutation tests are useful in power estimation.

Solution: They provide a non-parametric way to generate null distributions and assess significance without normality assumptions.

- (d) (2 points) What does a power curve illustrate?

Solution: How statistical power changes as a function of sample size for a fixed effect size and significance level.