



Pen-and-Paper Brain Teasers: Math for Econometrics

Easy Exercises (5-10 minutes each)

Linear Algebra - Vectors & Matrices

1. **Vector Addition** - Given vectors $\mathbf{a} = [2, -3, 5]$ and $\mathbf{b} = [1, 4, -2]$, compute $\mathbf{a} + \mathbf{b}$ and $\mathbf{a} - \mathbf{b}$.
2. **Dot Product** - Calculate the dot product of $\mathbf{u} = [3, 1, -2]$ and $\mathbf{v} = [4, -1, 5]$. Is the angle between them acute or obtuse?
3. **Matrix Multiplication (2x2)** - Multiply matrices:
$$A = \begin{bmatrix} 2 & 3 \\ 1 & 4 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 \\ -1 & 2 \end{bmatrix}$$
4. **Matrix Transpose** - Given $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$, find A^T and verify that $(A^T)^T = A$.
5. **Scalar Multiplication** - If $C = \begin{bmatrix} 2 & -1 \\ 0 & 3 \end{bmatrix}$, compute $3C$ and $-2C$.
6. **Diagonal Matrix** - Create a 3×3 diagonal matrix D with diagonal elements $[2, 5, -1]$. Compute D^2 .
7. **Identity Matrix** - Verify that $I_2 \times A = A$ where $I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ and $A = \begin{bmatrix} 3 & 7 \\ -2 & 4 \end{bmatrix}$.

Calculus - Derivatives

8. **Power Rule** - Find the derivative of $f(x) = 3x^4 - 2x^3 + 5x - 7$.
9. **Product Rule** - Differentiate $g(x) = (2x + 1)(x^2 - 3)$.
10. **Chain Rule (Basic)** - Find $\frac{d}{dx}(3x + 2)^5$.
11. **Exponential Function** - Compute $\frac{d}{dx}e^{2x}$ and $\frac{d}{dx}e^{-x}$.
12. **Natural Logarithm** - Find the derivative of $h(x) = \ln(x^2 + 1)$.

Calculus - Integrals

13. **Basic Integration** - Evaluate $\int (3x^2 + 4x - 1)dx$.
14. **Definite Integral** - Calculate $\int_0^2 (x^2 + 1)dx$.
15. **Exponential Integral** - Find $\int e^{3x}dx$.

Statistics - Descriptive Statistics

16. **Mean & Median** - Data: . Calculate mean and median. [\[1\]](#) [\[2\]](#) [\[3\]](#) [\[4\]](#) [\[5\]](#) [\[6\]](#) [\[7\]](#)
17. **Variance** - For data , compute the sample variance. [\[8\]](#) [\[9\]](#) [\[10\]](#) [\[11\]](#)
18. **Standard Deviation** - Given variance = 16, find standard deviation. If all values are multiplied by 2, what's the new standard deviation?
19. **Range & IQR** - Data: . Find range, Q1, Q3, and IQR. [\[2\]](#) [\[4\]](#) [\[10\]](#) [\[12\]](#) [\[13\]](#) [\[14\]](#) [\[15\]](#) [\[1\]](#)
20. **Z-score** - A test score is 85 with mean = 75 and $\sigma = 10$. Calculate the z-score. Is this score above or below average?

Probability Basics

21. **Basic Probability** - A bag has 3 red, 5 blue, and 2 green balls. What's P(red)? P(not blue)?
22. **Complementary Events** - If $P(\text{rain}) = 0.3$, what's $P(\text{no rain})$?
23. **Multiplication Rule** - Two independent coin flips. What's $P(\text{both heads})$?
24. **Expected Value (Discrete)** - A simple game: win \$10 with probability 0.3, lose \$5 with probability 0.7. What's the expected value?
25. **Variance of Distribution** - Random variable X : $P(X=1) = 0.5$, $P(X=3) = 0.5$. Find $E[X]$ and $\text{Var}(X)$.

Medium Exercises (10-30 minutes each)

Linear Algebra - Matrix Operations

1. **Determinant (2×2)** - Calculate the determinant of $A = \begin{bmatrix} 5 & 3 \\ 2 & 4 \end{bmatrix}$. Is A invertible?
2. **Matrix Inverse (2×2)** - Find the inverse of $B = \begin{bmatrix} 4 & 7 \\ 2 & 6 \end{bmatrix}$. Verify by computing $B \times B^{-1}$.
3. **System of Linear Equations** - Solve using substitution or elimination:
$$\begin{cases} 2x + y = 7 \\ x - y = 2 \end{cases}$$
4. **Matrix Form of System** - Write the system as $A\mathbf{x} = \mathbf{b}$ and solve using matrix inverse:
$$\begin{cases} 3x + 2y = 8 \\ x + y = 3 \end{cases}$$
5. **Eigenvalues (2×2)** - Find eigenvalues of $C = \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}$ by solving $\det(C - \lambda I) = 0$.
6. **Quadratic Form** - Given $Q = \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$ and $\mathbf{x} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, compute $\mathbf{x}^T Q \mathbf{x}$.
7. **Projection Matrix** - For vector $\mathbf{v} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$, find the projection onto $\mathbf{u} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.

Calculus - Optimization

8. **Critical Points** - Find critical points of $f(x) = x^3 - 6x^2 + 9x + 1$. Classify them as max, min, or inflection.
9. **Maxima/Minima** - A farmer has 100m of fence. What rectangular dimensions maximize area?
10. **Second Derivative Test** - For $g(x) = 2x^3 - 3x^2 - 12x + 5$, find local maxima and minima using the second derivative test.
11. **Partial Derivatives** - Find $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ where $f(x, y) = x^2y + 3xy^2 - 4$.
12. **Chain Rule (Multivariate)** - If $z = x^2 + y^2$, $x = 2t$, $y = 3t$, find $\frac{dz}{dt}$.
13. **Gradient Vector** - For $f(x, y) = 3x^2 + 2xy + y^2$, compute ∇f at point (1, 2).

Statistics - Distributions

14. **Normal Distribution** - $X \sim N(100, 15)$. Using z-table, find $P(X < 115)$ and $P(85 < X < 115)$.
15. **Standardization** - Convert $X \sim N(50, 10)$ to standard normal. What's the z-score for $X = 65$?
16. **Binomial Probability** - Flip fair coin 5 times. Calculate $P(\text{exactly 3 heads})$ and $P(\text{at least 3 heads})$.
17. **Covariance** - Given paired data: $X =$, $Y =$. Calculate $\text{Cov}(X, Y)$. [\[12\]](#) [\[16\]](#) [\[17\]](#) [\[18\]](#) [\[19\]](#) [\[8\]](#)
18. **Correlation** - From the covariance above, calculate the correlation coefficient $\rho(X, Y)$.
19. **Law of Total Probability** - Events: A_1 (disease, 1%), A_2 (no disease, 99%). Test: $P(+|\text{disease}) = 0.95$, $P(+|\text{no disease}) = 0.05$. Find $P(+)$.
20. **Bayes' Theorem** - Using the above, find $P(\text{disease}|+)$.

Econometrics Basics

21. **Simple Regression (By Hand)** - Data: $(x, y) = \{(1, 2), (2, 4), (3, 5), (4, 7)\}$. Calculate slope β_1 and intercept β_0 using formulas:
$$\beta_1 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$
22. **Fitted Values & Residuals** - Using regression from above, calculate fitted values \hat{y}_i and residuals $e_i = y_i - \hat{y}_i$.
23. **R-squared** - Calculate $R^2 = 1 - \frac{SS_{res}}{SS_{tot}}$ where $SS_{res} = \sum e_i^2$ and $SS_{tot} = \sum (y_i - \bar{y})^2$.
24. **Elasticity** - If demand function is $Q = 100 - 2P$, calculate price elasticity at $P = 20$.
25. **Log Returns** - Stock prices: Day 1: \$100, Day 2: \$105, Day 3: \$103. Calculate log returns for each day.

Hard Exercises (1-2 hours each)

Linear Algebra - Advanced

- 1. Eigenvalue Decomposition** - For matrix $A = \begin{bmatrix} 4 & 1 \\ 2 & 3 \end{bmatrix}$:
 - Find eigenvalues λ_1, λ_2
 - Find corresponding eigenvectors
 - Write A as $A = PDP^{-1}$ where D is diagonal
 - Verify by computing PDP^{-1}
- 2. Least Squares (Matrix Method)** - Data points: (1,1), (2,3), (3,2), (4,5). Fit line $y = \beta_0 + \beta_1 x$ using normal equations:
 $(X^T X)\beta = X^T y$
Set up matrices, compute $(X^T X)^{-1}$, solve for β .
- 3. Positive Definite Matrix** - Prove that $A = \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix}$ is positive definite by:
 - Checking all eigenvalues > 0
 - Checking leading principal minors > 0
 - Showing $\mathbf{x}^T A \mathbf{x} > 0$ for any non-zero \mathbf{x}
- 4. Diagonalization Application** - Given $A = \begin{bmatrix} 3 & 1 \\ 0 & 2 \end{bmatrix}$, compute A^{10} using diagonalization (much easier than multiplying A by itself 10 times).
- 5. Singular Value Decomposition (2×2)** - For $B = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$, find the SVD: $B = U\Sigma V^T$. (Hint: this is already diagonal!)

Calculus - Multivariate Optimization

- 6. Unconstrained Optimization** - Find the minimum of $f(x, y) = x^2 + 2y^2 - 2xy + 2x - 4y + 5$. Use:
 - Find critical points (set $\nabla f = 0$)
 - Compute Hessian matrix
 - Use second derivative test (check if Hessian is positive definite)
- 7. Lagrange Multipliers** - Maximize $f(x, y) = xy$ subject to $x + y = 10$. Set up Lagrangian, find critical points, solve for x, y, λ .
- 8. Constrained Optimization (Inequality)** - Minimize $f(x, y) = x^2 + y^2$ subject to $x + 2y \geq 4$. Use KKT conditions or graphical analysis.
- 9. Taylor Approximation** - Approximate $e^{0.1}$ using Taylor series up to 3rd degree term. Calculate error compared to actual value.
- 10. Integration by Parts** - Evaluate $\int x e^x dx$ using integration by parts formula.

Statistics - Hypothesis Testing

11. **T-test (By Hand)** - Sample 1: (n=5). Sample 2: (n=5). ^{[4] [5] [6] [7] [11] [20] [21] [22] [1] [2]}
- Calculate means and pooled variance
 - Compute t-statistic
 - With df=8, critical value (95%) ≈ 2.306 . Reject $H_0: \mu_1 = \mu_2$?
12. **Chi-Square Test** - Observed frequencies:.. Expected (all equal):.. Calculate χ^2 statistic: ^{[6] [14] [15]}
- $$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$
- With df=3, critical value (95%) ≈ 7.815 . Significant?
13. **Confidence Interval** - Sample mean = 50, s = 8, n = 16. Construct 95% CI for population mean (use t = 2.131 for df=15).
14. **Type I and Type II Errors** - A court trial analogy: Define null hypothesis H_0 (defendant innocent). What is Type I error? Type II error? Which is worse in this context?
15. **Power Calculation** - For testing $H_0: \mu = 100$ vs $H_1: \mu = 105$, with $\sigma = 10$, n = 25, $\alpha = 0.05$. Calculate the power (probability of rejecting H_0 when $\mu = 105$).

Econometrics - Regression Analysis

16. **Multiple Regression Setup** - You have data on house prices (Y), square footage (X_1), and number of bedrooms (X_2). Write out:
- The regression model: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon$
 - Matrix form: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$
 - Normal equations: $(\mathbf{X}^T \mathbf{X})\boldsymbol{\beta} = \mathbf{X}^T \mathbf{y}$
 - For small dataset $\{(Y, X_1, X_2)\}$: (200, 1000, 2), (250, 1200, 3), (300, 1500, 3), set up and solve
17. **Omitted Variable Bias** - True model: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon$. You mistakenly estimate: $Y = \alpha_0 + \alpha_1 X_1 + u$.
- Show that α_1 is biased if X_1 and X_2 are correlated
 - Derive the bias formula: $\text{Bias} = \beta_2 \times \frac{\text{Cov}(X_1, X_2)}{\text{Var}(X_1)}$
 - Give economic interpretation
18. **Heteroskedasticity Test** - Given residuals from regression: [2, -1, 4, -3, 5, -2, 6, -4]. Plot e_i^2 against fitted values. Does variance appear constant? What problem does this indicate?
19. **Autocorrelation (Durbin-Watson)** - Residuals (in time order): [1, 2, -1, -2, 0, 1, -1, 2]. Calculate Durbin-Watson statistic:
- $$DW = \frac{\sum_{t=2}^n (e_t - e_{t-1})^2}{\sum_{t=1}^n e_t^2}$$
- Is there evidence of autocorrelation? (DW ≈ 2 means no autocorrelation)
20. **Instrumental Variables** - Regression: $\text{Education} = \beta_0 + \beta_1 \text{Income} + \epsilon$. Problem: Income is endogenous (reverse causality). Propose valid instrument (e.g., parental education). What conditions must instrument satisfy?

Advanced Exercises (2-4 hours each)

Linear Algebra - Higher Dimensions

1. **3×3 Determinant & Inverse** - Calculate the determinant of matrix A using cofactor expansion:

$$A = \begin{bmatrix} 2 & 1 & 3 \\ 0 & 4 & 1 \\ 1 & 2 & 2 \end{bmatrix}$$

Then find A^{-1} using the adjugate method: $A^{-1} = \frac{1}{\det A} \text{adj}(A)$. Verify your answer.

2. **4D Vector Operations** - Given four-dimensional vectors:

- $\mathbf{a} = [1, -2, 3, 4]$, $\mathbf{b} = [2, 1, -1, 2]$, $\mathbf{c} = [0, 3, 2, -1]$
- Calculate: $\mathbf{a} \cdot \mathbf{b}$, $\|\mathbf{a}\|$, angle between \mathbf{a} and \mathbf{b}
- Find vector \mathbf{d} orthogonal to both \mathbf{a} and \mathbf{b} (use Gram-Schmidt on 4D space)
- Compute projection of \mathbf{c} onto \mathbf{a}

3. **3×3 Eigenvalue Problem** - For matrix:

$$B = \begin{bmatrix} 6 & -2 & 2 \\ -2 & 3 & -1 \\ 2 & -1 & 3 \end{bmatrix}$$

- Find all three eigenvalues by solving the characteristic polynomial (cubic equation)
- Find corresponding eigenvectors for each eigenvalue
- Verify orthogonality of eigenvectors (B is symmetric)

4. **3×3 System of Equations** - Solve using Gaussian elimination with back-substitution:

$$\begin{cases} 2x + y - z = 8 \\ -3x - y + 2z = -11 \\ -2x + y + 2z = -3 \end{cases}$$

Show all row operations step-by-step to reach row echelon form.

5. **Gram-Schmidt Orthogonalization (3D)** - Given basis vectors:

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \quad \mathbf{v}_3 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$

Apply Gram-Schmidt to produce orthonormal basis $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$. Verify orthonormality.

6. **QR Decomposition (3×3)** - Using the result from exercise 5, write the original matrix with columns $[\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3]$ as $A = QR$ where Q is orthogonal and R is upper triangular.

Calculus - Higher-Order Polynomials

7. **Fifth-Degree Polynomial** - Find all critical points and classify them for:

$$f(x) = x^5 - 5x^4 + 5x^3 + 5x^2 - 6x$$

Use first and second derivative tests. Identify intervals of increase/decrease and concavity.

8. **Three-Variable Optimization** - Find and classify critical points of:

$$f(x, y, z) = x^2 + y^2 + z^2 - xy - xz - yz$$

- Compute gradient and set to zero
- Compute 3×3 Hessian matrix
- Use eigenvalue test for classification

9. **Multivariate Taylor Series** - Find the second-order Taylor expansion of $f(x, y) = e^{x+y}$ around point (0,0):

$$f(x, y) \approx f(0, 0) + \nabla f \cdot \begin{bmatrix} x \\ y \end{bmatrix} + \frac{1}{2} \begin{bmatrix} x & y \end{bmatrix} H \begin{bmatrix} x \\ y \end{bmatrix}$$

where H is the Hessian. Evaluate at (0.1, 0.2) and compare with actual value.

10. **Lagrange Multipliers (3 Variables)** - Minimize $f(x, y, z) = x^2 + y^2 + z^2$ subject to constraint $x + 2y + 3z = 6$.

- Set up Lagrangian with one constraint
- Solve system of 4 equations in 4 unknowns
- Verify solution using geometric interpretation (closest point on plane to origin)

11. **Polynomial Long Division** - Divide $p(x) = 2x^4 + 3x^3 - x^2 + 5x - 2$ by $q(x) = x^2 + x - 1$.

Find quotient and remainder, verify $p(x) = q(x) \cdot \text{quotient} + \text{remainder}$.

12. **Partial Fraction Decomposition** - Decompose the rational function:

$$\frac{2x^3 + 3x^2 + 4x + 5}{(x+1)(x-1)(x^2+1)}$$

into partial fractions. Then integrate the result.

Statistics - Multivariate Analysis

13. **3×3 Covariance Matrix** - Given data for three variables X, Y, Z (5 observations each):

- X: $\begin{bmatrix} 16 \\ 17 \\ 18 \\ 8 \\ 12 \end{bmatrix}$
- Y: $\begin{bmatrix} 17 \\ 8 \\ 12 \end{bmatrix}$
- Z: $\begin{bmatrix} 9 \\ 18 \\ 8 \\ 12 \end{bmatrix}$

Calculate the 3×3 covariance matrix and 3×3 correlation matrix by hand.

14. **Multinomial Distribution** - A die is rolled 12 times. Calculate the probability of getting exactly 2 ones, 3 twos, 3 threes, 2 fours, 1 five, and 1 six using the multinomial formula:

$$P = \frac{n!}{n_1! n_2! \dots n_k!} p_1^{n_1} p_2^{n_2} \dots p_k^{n_k}$$

15. **Principal Component Analysis (By Hand)** - Using the 3×3 covariance matrix from exercise 13:

- Find eigenvalues and eigenvectors
- Order eigenvectors by eigenvalue magnitude (these are principal components)
- Calculate proportion of variance explained by first two components
- Transform one data point into PC space

Econometrics - Advanced Topics

16. Three-Variable Regression (Matrix Form) - Fit model

$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \epsilon$ to data:

Y	X ₁	X ₂	X ₃
10	1	2	1
15	2	3	2
12	1	4	1
18	3	5	2
20	4	6	3

- Construct design matrix X (5×4, including intercept column)
- Compute $X^T X$ (4×4 matrix)
- Compute $(X^T X)^{-1}$ (requires 4×4 inversion)
- Solve for $\beta = (X^T X)^{-1} X^T y$
- Calculate fitted values and R^2

17. F-Test for Joint Significance - Using regression from exercise 16:

- Test $H_0: \beta_2 = \beta_3 = 0$ (last two coefficients jointly zero)
- Calculate $SSR_{\text{restricted}}$ (regress Y on X_1 only) and $SSR_{\text{unrestricted}}$ (full model)
- Compute F-statistic: $F = \frac{(SSR_r - SSR_u)/q}{SSR_u/(n-k)}$
- Compare with critical value $F(2, n-4)$ at 5% level

18. Polynomial Regression - Fit cubic model $Y = \beta_0 + \beta_1 X + \beta_2 X^2 + \beta_3 X^3 + \epsilon$ to data:

- (X, Y) : (1, 2), (2, 3), (3, 8), (4, 15), (5, 27)
- Set up design matrix with columns $[1, X, X^2, X^3]$
- Solve normal equations
- Evaluate model at $X = 3.5$

19. Variance-Covariance Matrix of Estimators - For the regression in exercise 16:

- Estimate error variance: $\hat{\sigma}^2 = \frac{SSR}{n-k}$
- Calculate $\text{Var}(\hat{\beta}) = \hat{\sigma}^2 (X^T X)^{-1}$
- Extract standard errors (square roots of diagonal elements)
- Compute t-statistics for each coefficient

20. Interaction Terms - Model with interaction:

$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 (X_1 \times X_2) + \epsilon$

Data:

Y	X ₁	X ₂
5	1	2

Y	X ₁	X ₂
10	2	3
8	1	4
15	3	3
12	2	4

- Create interaction variable $X_1 \times X_2$
- Solve for coefficients
- Interpret β_3 : how does the effect of X_1 on Y depend on X_2 ?

Solution Methods Reference (Check Yourself!)

Linear Algebra Methods

Easy Level:

1. **Vector Addition** - Component-wise addition/subtraction
2. **Dot Product** - Dot product formula: $u \cdot v = u_1v_1 + u_2v_2 + u_3v_3$; Sign test: positive \rightarrow acute, negative \rightarrow obtuse
3. **Matrix Multiplication (2×2)** - Row-by-column multiplication rule
4. **Matrix Transpose** - Swap rows and columns (element a_{ij} becomes a_{ji})
5. **Scalar Multiplication** - Multiply each element by the scalar
6. **Diagonal Matrix** - For diagonal matrices, $(D^2)_{ii} = (D_{ii})^2$, off-diagonal elements remain 0
7. **Identity Matrix** - Matrix multiplication; identity matrix property

Medium Level:

1. **Determinant (2×2)** - 2×2 determinant formula: $\det(A) = ad - bc$; Invertibility test: $\det(A) \neq 0$
2. **Matrix Inverse (2×2)** - 2×2 inverse formula: $A^{-1} = (1/\det(A)) \cdot [d, -b; -c, a]$ where $A = [a, b; c, d]$
3. **System of Linear Equations** - Substitution method; Elimination (addition) method; Graphical method
4. **Matrix Form of System** - Matrix equation: $x = A^{-1}b$; Cramer's rule using determinants
5. **Eigenvalues (2×2)** - Characteristic equation: $\det(A - \lambda I) = 0$, yields quadratic equation in λ
6. **Quadratic Form** - Matrix multiplication: compute Qx first, then $x^T(Qx)$
7. **Projection Matrix** - Vector projection formula: $\text{proj}_u(v) = [(v \cdot u)/(u \cdot u)]u$

Hard Level:

1. **Eigenvalue Decomposition** - Characteristic polynomial: $\det(A - \lambda I) = 0$; Eigenvector: solve $(A - \lambda I)v = 0$; Diagonalization: $P = [v_1 \ v_2]$, $D = \text{diag}(\lambda_1, \lambda_2)$

2. **Least Squares (Matrix Method)** - Normal equations: $X^T X \beta = X^T y$; Matrix solution: $\beta = (X^T X)^{-1} X^T y$; QR decomposition (advanced); Geometric interpretation: projection onto column space of X
3. **Positive Definite Matrix** - Eigenvalue test; Sylvester's criterion (principal minors); Quadratic form test; Cholesky decomposition (advanced)
4. **Diagonalization Application** - Diagonalization: $A = P D P^{-1}$, then $A^n = P D^n P^{-1}$; For diagonal D , D^n is easy (just raise diagonal elements to power n)
5. **Singular Value Decomposition (2x2)** - SVD algorithm: compute eigenvalues of $B^T B$ and $B B^T$; For diagonal matrices, SVD is trivial; Relationship to eigenvalue decomposition

Advanced Level:

1. **3x3 Determinant & Inverse** - Cofactor expansion (Laplace expansion) along any row/column; Rule of Sarrus (for 3x3 only); Adjugate method: $A^{-1} = \text{adj}(A)/\det(A)$; Gauss-Jordan elimination
2. **4D Vector Operations** - Standard dot product in n dimensions; Magnitude: $\|v\| = \sqrt{v \cdot v}$; Angle: $\cos(\theta) = (a \cdot b)/(\|a\| \|b\|)$; Gram-Schmidt process for orthogonalization
3. **3x3 Eigenvalue Problem** - Characteristic polynomial is cubic; Cardano's formula or numerical approximation; For symmetric matrices: real eigenvalues, orthogonal eigenvectors (Spectral theorem)
4. **3x3 System of Equations** - Gaussian elimination; Row echelon form; Back substitution; Gauss-Jordan for reduced row echelon form; LU decomposition
5. **Gram-Schmidt Orthogonalization (3D)** - Gram-Schmidt algorithm: $u_1 = v_1/\|v_1\|$, $u_2 = (v_2 - \text{proj}_{\{u_1\}} v_2)/\|v_2 - \text{proj}_{\{u_1\}} v_2\|$, etc.; QR decomposition is built on this
6. **QR Decomposition (3x3)** - $A = QR$ where Q is orthogonal (Q from Gram-Schmidt), R is upper triangular (inner products)

Calculus Methods

Easy Level:

8. **Power Rule** - Power rule: $d/dx(x^n) = nx^{n-1}$; Linearity of derivatives
9. **Product Rule** - Product rule: $(uv)' = u'v + uv'$; Alternative: expand first then differentiate
10. **Chain Rule (Basic)** - Chain rule: $d/dx[f(g(x))] = f'(g(x)) \cdot g'(x)$
11. **Exponential Function** - Exponential rule with chain rule: $d/dx(e^{kx}) = ke^{kx}$
12. **Natural Logarithm** - Logarithmic differentiation with chain rule: $d/dx[\ln(u)] = u'/u$
13. **Basic Integration** - Power rule for integration: $\int x^n dx = x^{n+1}/(n+1) + C$; Linearity
14. **Definite Integral** - Fundamental Theorem of Calculus: $\int_a^b f(x) dx = F(b) - F(a)$
15. **Exponential Integral** - Exponential integration rule: $\int e^{kx} dx = (1/k)e^{kx} + C$

Medium Level:

8. **Critical Points** - First derivative test: set $f'(x) = 0$; Second derivative test: $f''(x) > 0$ (min), $f''(x) < 0$ (max)
9. **Maxima/Minima** - Constraint optimization: express area $A(x)$ subject to perimeter constraint; Calculus: find $dA/dx = 0$
10. **Second Derivative Test** - Second derivative test: compute $g'(x) = 0$ for critical points, then

evaluate $g''(x)$ at each

11. **Partial Derivatives** - Partial differentiation: treat other variables as constants

12. **Chain Rule (Multivariate)** - Substitution then differentiate; Multivariate chain rule: $dz/dt = (\partial z/\partial x)(dx/dt) + (\partial z/\partial y)(dy/dt)$

13. **Gradient Vector** - Gradient vector: $\nabla f = [\partial f/\partial x, \partial f/\partial y]$

Hard Level:

6. **Unconstrained Optimization** - First-order condition: $\nabla f = 0$ (system of equations); Hessian matrix $H = [\partial^2 f/\partial x_i \partial x_j]$; Second-order condition: check eigenvalues of H or determinant tests; Bordered Hessian (for constrained)

7. **Lagrange Multipliers** - Lagrangian: $L(x, y, \lambda) = f(x, y) - \lambda(g(x, y) - c)$; First-order conditions: $\nabla L = 0$; Geometric interpretation: gradient of f parallel to gradient of constraint

8. **Constrained Optimization (Inequality)** - KKT (Karush-Kuhn-Tucker) conditions for inequality constraints; Complementary slackness; Graphical method: sketch feasible region; Check boundary vs interior solutions

9. **Taylor Approximation** - Taylor series expansion: $f(x) \approx f(a) + f'(a)(x-a) + f''(a)(x-a)^2/2! + \dots$; Remainder term (Lagrange form) for error estimation

10. **Integration by Parts** - Integration by parts: $\int u dv = uv - \int v du$; Choice of u and dv is critical (LIATE rule: Logarithmic, Inverse trig, Algebraic, Trig, Exponential)

Advanced Level:

7. **Fifth-Degree Polynomial** - Derivative of polynomial is one degree lower; Rational root theorem for finding roots; Synthetic division; Newton's method for numerical solutions

8. **Three-Variable Optimization** - Gradient in 3D: $\nabla f = [\partial f/\partial x, \partial f/\partial y, \partial f/\partial z]$; 3×3 Hessian matrix; Eigenvalue test for definiteness; Sylvester's criterion

9. **Multivariate Taylor Series** - Multivariate Taylor: $f(x) \approx f(a) + \nabla f(a) \cdot (x-a) + \frac{1}{2}(x-a)^T H(a)(x-a)$; Hessian contains all second partials

10. **Lagrange Multipliers (3 Variables)** - Lagrangian: $L = f - \lambda g$; $\nabla L = 0$ gives $n+1$ equations; Geometric: level sets of f tangent to constraint surface

11. **Polynomial Long Division** - Polynomial long division algorithm; Remainder theorem; Factor theorem

12. **Partial Fraction Decomposition** - Method of partial fractions; Cover-up method; Heaviside's method; Integration of rational functions

Statistics Methods

Easy Level:

16. **Mean & Median** - Mean = sum/n ; Median = middle value after sorting (for odd n)

17. **Variance** - Sample variance formula: $s^2 = \sum (x_i - \bar{x})^2 / (n-1)$

18. **Standard Deviation** - $SD = \sqrt{\text{variance}}$; Scaling property: $SD(kX) = |k| \cdot SD(X)$

19. **Range & IQR** - Range = $\text{max} - \text{min}$; Quartiles: $Q1 = 25\text{th percentile}$, $Q3 = 75\text{th percentile}$; $IQR = Q3 - Q1$

20. **Z-score** - Z-score standardization: $z = (x - \mu)/\sigma$

21. **Basic Probability** - Classical probability: $P(A) = \text{favorable}/\text{total}$; Complement rule: $P(\text{not } A) = 1 - P(A)$

22. **Complementary Events** - Complement rule: $P(A^c) = 1 - P(A)$

23. **Multiplication Rule** - Multiplication rule for independent events: $P(A \cap B) = P(A) \cdot P(B)$

24. **Expected Value (Discrete)** - Expected value formula: $E[X] = \sum x_i \cdot P(x_i)$

25. **Variance of Distribution** - $E[X] = \sum x_i \cdot P(x_i)$; $\text{Var}(X) = E[X^2] - (E[X])^2$

Medium Level:

14. **Normal Distribution** - Z-score transformation: $z = (x - \mu)/\sigma$; Standard normal table lookup; Symmetry properties of normal distribution

15. **Standardization** - Standardization formula: $Z = (X - \mu)/\sigma$, where $Z \sim N(0, 1)$

16. **Binomial Probability** - Binomial probability formula: $P(X=k) = C(n,k) \cdot p^k \cdot (1-p)^{(n-k)}$; Binomial coefficient: $C(n,k) = n!/(k!(n-k)!)$

17. **Covariance** - Sample covariance formula: $\text{Cov}(X, Y) = \sum (x_i - \bar{x})(y_i - \bar{y})/(n-1)$

18. **Correlation** - Pearson correlation: $\rho = \text{Cov}(X, Y)/(\text{SD}(X) \cdot \text{SD}(Y))$

19. **Law of Total Probability** - Law of total probability: $P(B) = \sum P(B|A_i)P(A_i)$

20. **Bayes' Theorem** - Bayes' theorem: $P(A|B) = P(B|A)P(A)/P(B)$

Hard Level:

11. **T-test (By Hand)** - Two-sample t-test with pooled variance: $t = (\bar{x}_1 - \bar{x}_2)/\sqrt{(s_p^2(1/n_1 + 1/n_2))}$;

Pooled variance: $s_p^2 = [(n_1-1)s_1^2 + (n_2-1)s_2^2]/(n_1+n_2-2)$; Welch's t-test (unequal variances);

Paired t-test (if data is paired)

12. **Chi-Square Test** - Chi-square goodness-of-fit test; Degrees of freedom: $df = k - 1 - p$ (k categories, p estimated parameters); Chi-square test of independence (for contingency tables)

13. **Confidence Interval** - t-distribution CI: $\bar{x} \pm t \cdot (s/\sqrt{n})$; Z-distribution CI (for large n or known σ); Bootstrap CI (computational method)

14. **Type I and Type II Errors** - Type I error (α): reject true H_0 (false positive); Type II error (β): fail to reject false H_0 (false negative); Power = $1 - \beta$; Trade-off between errors

15. **Power Calculation** - Power = $P(\text{reject } H_0 | H_1 \text{ true})$; Calculate critical value under H_0 ; Find probability of exceeding critical value under H_1 ; Use z-distribution or t-distribution

Advanced Level:

13. **3x3 Covariance Matrix** - Covariance matrix: $\Sigma_{ij} = \text{Cov}(X_i, X_j)$; Symmetric matrix; Diagonal contains variances; Correlation matrix: standardize by dividing by standard deviations

14. **Multinomial Distribution** - Multinomial probability: $P = n!/(n_1!n_2!\dots n_k!) \cdot p_1^{n_1} \cdot p_2^{n_2} \cdot \dots \cdot p_k^{n_k}$; Generalization of binomial

15. **Principal Component Analysis (By Hand)** - PCA steps: (1) Compute covariance matrix, (2) Find eigenvalues/eigenvectors, (3) Sort by eigenvalue, (4) Transform data using eigenvector matrix; Variance explained = eigenvalue/sum of eigenvalues

Econometrics Methods

Medium Level:

21. **Simple Regression (By Hand)** - OLS (Ordinary Least Squares) formulas for β_1 and β_0 ; Method of moments; Geometric interpretation: minimize sum of squared residuals

22. **Fitted Values & Residuals** - Fitted value formula: $\hat{y}_i = \beta_0 + \beta_1 x_i$; Residual: $e_i = y_i - \hat{y}_i$

23. **R-squared** - R^2 formula using sum of squares; Alternative: $R^2 = (\text{correlation coefficient})^2$

24. **Elasticity** - Point elasticity formula: $\epsilon = (dQ/dP) \cdot (P/Q)$

25. **Log Returns** - Log return formula: $r_t = \ln(P_t/P_{t-1})$

Hard Level:

16. **Multiple Regression Setup** - OLS in matrix form; Interpretation of coefficients (partial

derivatives); Multicollinearity concerns; Adjusted R^2

17. **Omitted Variable Bias** - Omitted variable bias formula derivation; Auxiliary regression: regress X_2 on X_1 ; Direction of bias depends on sign of β_2 and correlation; Frisch-Waugh-Lovell theorem

18. **Heteroskedasticity Test** - Visual inspection: plot residuals vs fitted values or vs X ; Breusch-Pagan test; White test; Goldfeld-Quandt test; Consequences: OLS still unbiased but inefficient, standard errors wrong

19. **Autocorrelation (Durbin-Watson)** - Durbin-Watson test: $DW \approx 2(1 - \rho)$ where ρ is first-order autocorrelation; Breusch-Godfrey test (higher-order); Ljung-Box test; Consequences: OLS inefficient, standard errors biased

20. **Instrumental Variables** - Two-Stage Least Squares (2SLS); Instrument relevance: $\text{Cov}(Z, X) \neq 0$; Instrument exogeneity: $\text{Cov}(Z, \varepsilon) = 0$; First-stage regression: regress X on Z ; Second-stage: regress Y on \hat{X} ; Weak instruments problem; Overidentification tests (if multiple instruments)

Advanced Level:

16. **Three-Variable Regression (Matrix Form)** - Design matrix construction (include column of 1s for intercept); $X^T X$ is $k \times k$ symmetric matrix; $(X^T X)^{-1}$ requires matrix inversion; $\hat{\beta} = (X^T X)^{-1} X^T y$; $R^2 = 1 - SSR/SST$

17. **F-Test for Joint Significance** - F-test for multiple restrictions; F-statistic: $F = [(SSR_R - SSR_U)/q]/[SSR_U/(n-k)]$; q = number of restrictions; F-distribution with $(q, n-k)$ degrees of freedom; Relationship to R^2

18. **Polynomial Regression** - Polynomial regression as special case of multiple regression; Design matrix includes powers of X ; Multicollinearity issues with high-degree polynomials; Centered polynomials reduce collinearity

19. **Variance-Covariance Matrix of Estimators** - $\text{Var}(\hat{\beta}) = \sigma^2 (X^T X)^{-1}$; Estimate σ^2 : residual variance $s^2 = SSR/(n-k)$; Standard errors: $SE(\hat{\beta}_j) = \sqrt{[(X^T X)^{-1}]_{jj}} \cdot s$; t-statistics for hypothesis testing

20. **Interaction Terms** - Interaction: effect of X_1 depends on level of X_2 ; Coefficient interpretation: $\partial Y / \partial X_1 = \beta_1 + \beta_3 X_2$; Centering variables before creating interactions; Marginal effects at different levels

Research Keywords by Topic

Linear Algebra

- **Basic operations:** Vector spaces, linear combinations, span, linear independence
- **Matrix theory:** Rank, nullity, row space, column space, null space
- **Determinants:** Properties, cofactor expansion, Cramer's rule
- **Eigentheory:** Characteristic polynomial, algebraic/geometric multiplicity, diagonalization, Jordan normal form
- **Orthogonality:** Inner products, orthogonal projections, Gram-Schmidt, QR decomposition
- **Spectral theory:** Spectral theorem, positive definiteness, Cholesky decomposition
- **SVD:** Singular value decomposition, pseudoinverse, low-rank approximation, matrix norms

- **Advanced:** Tensor products, exterior algebra, matrix exponential, Cayley-Hamilton theorem

Calculus

- **Single variable:** Limits, continuity, mean value theorem, L'Hôpital's rule
- **Differentiation:** Power rule, product rule, quotient rule, chain rule, implicit differentiation, logarithmic differentiation
- **Integration:** Substitution, integration by parts, partial fractions, trigonometric substitution, improper integrals
- **Series:** Taylor series, Maclaurin series, convergence tests, power series
- **Multivariate:** Partial derivatives, gradient, directional derivatives, Jacobian matrix, implicit function theorem
- **Optimization:** Lagrange multipliers, KKT conditions, convexity, Jensen's inequality, envelope theorem
- **Vector calculus:** Divergence, curl, line integrals, Green's theorem, Stokes' theorem
- **Advanced:** Differential forms, calculus of variations, Legendre transform, Hamiltonian mechanics

Statistics & Probability

- **Foundations:** Sample space, events, probability axioms, conditional probability
- **Distributions:** Discrete (binomial, Poisson, geometric), Continuous (normal, exponential, chi-square, t, F)
- **Central Limit Theorem:** Sampling distributions, standard error, law of large numbers
- **Estimation:** Point estimation, MLE (Maximum Likelihood), method of moments, bias-variance tradeoff
- **Hypothesis testing:** Null hypothesis, p-value, Type I/II errors, power, multiple testing
- **Confidence intervals:** Coverage probability, pivotal quantities, bootstrap
- **Regression:** Least squares, Gauss-Markov theorem, residual analysis
- **Multivariate:** Joint distributions, marginal distributions, independence, copulas
- **Advanced:** Sufficient statistics, Rao-Blackwell theorem, exponential families, decision theory

Econometrics

- **OLS foundations:** BLUE, Gauss-Markov theorem, assumptions (linearity, exogeneity, homoskedasticity, no autocorrelation)
- **Violations & remedies:**
 - Heteroskedasticity: White standard errors, WLS (Weighted Least Squares), FGLS
 - Autocorrelation: Newey-West standard errors, Cochrane-Orcutt, AR models
 - Multicollinearity: VIF, ridge regression, PCA

- Endogeneity: IV (Instrumental Variables), 2SLS, GMM (Generalized Method of Moments)
- **Model selection:** AIC, BIC, adjusted R^2 , cross-validation, information criteria
- **Causality:** Randomized experiments, natural experiments, difference-in-differences, regression discontinuity, matching, propensity scores
- **Time series:** Stationarity, unit root tests (ADF, PP), ARIMA, VAR, cointegration, ECM (Error Correction Model), GARCH
- **Panel data:** Fixed effects, random effects, Hausman test, clustered standard errors, difference-in-differences
- **Limited dependent variables:** Logit, probit, Tobit, selection models (Heckman), count data (Poisson, negative binomial)
- **Advanced:** Quantile regression, nonparametric regression, structural models, dynamic models, spatial econometrics

Numerical & Computational Methods

- **Linear systems:** Gaussian elimination, LU decomposition, iterative methods (Jacobi, Gauss-Seidel)
- **Eigenproblems:** Power iteration, QR algorithm, Lanczos method
- **Optimization:** Gradient descent, Newton-Raphson, quasi-Newton (BFGS), conjugate gradient, simplex method
- **Root finding:** Bisection, Newton's method, secant method
- **Integration:** Trapezoidal rule, Simpson's rule, Gaussian quadrature, Monte Carlo
- **Differential equations:** Euler's method, Runge-Kutta, finite differences

This comprehensive reference guide provides all the mathematical methods and research directions needed to tackle exercises at every level!



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22. <https://mojoauth.com/encryption-decryption/cast-256-224-encryption--javascript-in-browser/>