

Discovery Worksheet

Introduction to Machine Learning & AI

Discover the Patterns Before the Formulas

Machine Learning for Smarter Innovation

Week 0: Pre-Lecture Activity

How to Use This Worksheet

Important: Complete this **BEFORE** attending the lecture!

Your mission:

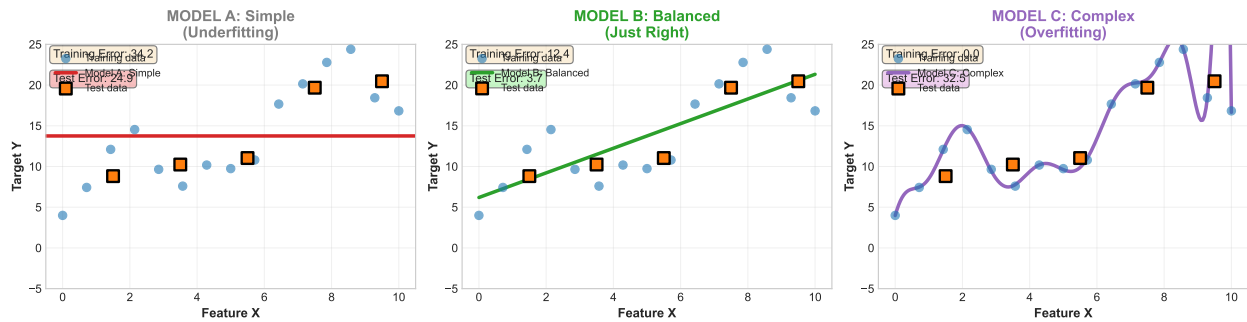
- Look at the 6 charts (one per section)
- Answer the discovery questions using calculations and observations
- Find the patterns YOURSELF before hearing the technical terms
- Compare answers with a partner (different answers are OK!)

Time needed: 75-90 minutes

Materials: Calculator, ruler, pencil

Learning Philosophy: You already understand these concepts from everyday life. This worksheet helps you discover the mathematical patterns behind what you already know intuitively. The lecture will then give you the formal language to describe your discoveries.

Discovery 1: The Overfitting Paradox



Observe the Chart

Three models are trying to predict house prices from square footage. All three were trained on the same blue dots (training data). Orange squares are NEW houses the models have never seen (test data).

Discovery Tasks

Task 1: Calculate Training Errors

For each model, measure how far predictions are from the blue training points.

- Model A (red horizontal line): Approx. training error = _____
- Model B (green curve): Approx. training error = _____
- Model C (purple wiggly): Approx. training error = _____

Which model has the LOWEST training error? _____

Task 2: Calculate Test Errors

Now measure how well each model predicts the NEW orange test points.

- Model A test error: _____
- Model B test error: _____
- Model C test error: _____

Which model has the LOWEST test error? _____

Task 3: The Paradox

Discovery Question:

Model C has the BEST training performance but the WORST test performance!
Why does perfect training lead to poor testing?

Task 4: Plot the Trade-off

On the grid below, plot training error (x-axis) vs test error (y-axis) for all three models:



What pattern do you see? _____

Task 5: Predict Model D

Imagine Model D is even MORE complex than Model C (degree 20 polynomial).

- Predicted training error: _____ (Why? _____)
- Predicted test error: _____ (Why? _____)

Key Insight:

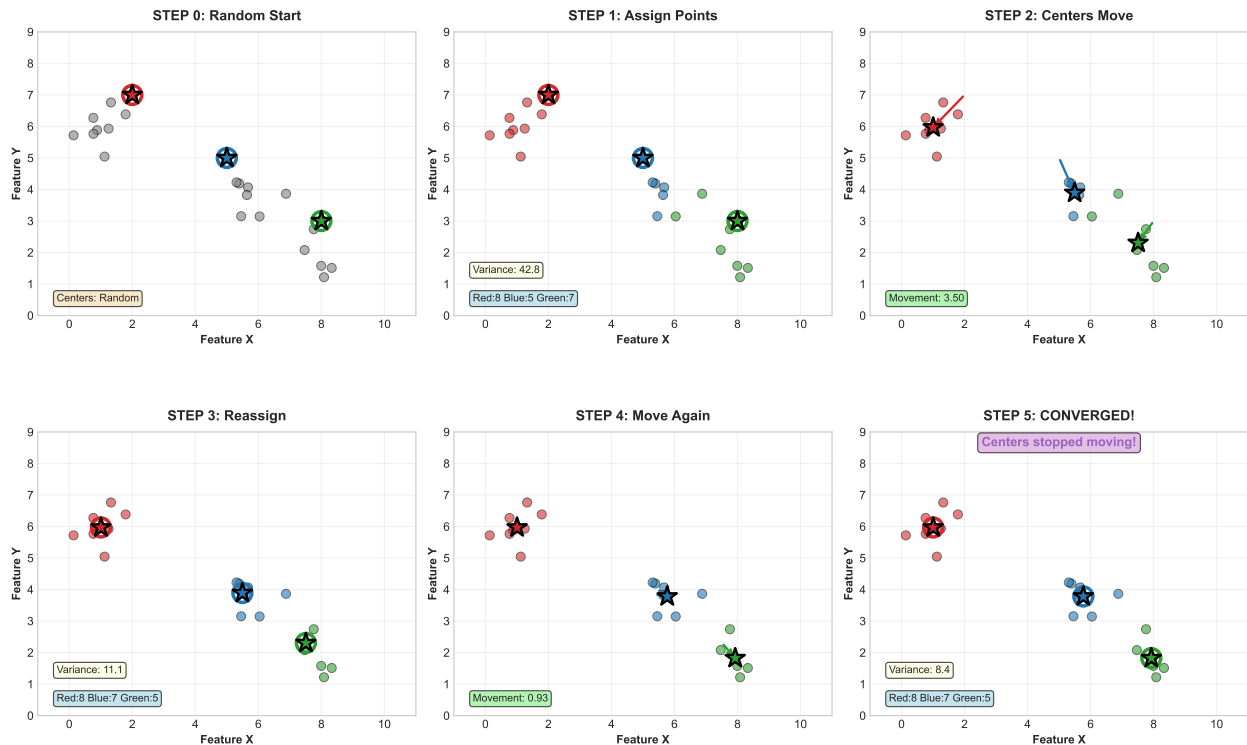
Model A problem: _____ (too simple)

Model C problem: _____ (too complex)

Model B achieves: _____ (balanced)

In the lecture, you'll learn these have names: Bias and Variance

Discovery 2: The Moving Centers Algorithm



Observe the Chart

Watch how three centers (marked with stars) move over 5 steps to find natural groups in the data.

Discovery Tasks

Task 1: Calculate Center Movements

In Step 2, centers move to the average position of their assigned points.

Calculate the Red center position in Step 2:

- Red cluster points visible in Step 1: Count = _____
- Approximate center coordinates: (_____, _____)
- Movement distance from Step 0 to Step 2: _____ units

Task 2: Calculate Within-Cluster Variance

Variance measures how spread out points are from their center.

For the Red cluster in Step 1:

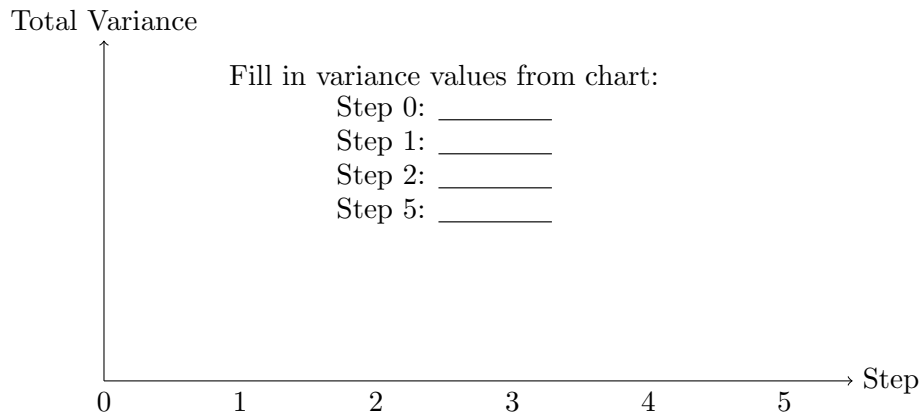
$$\text{Variance} = \frac{1}{n} \sum (\text{distance from point to center})^2$$

Pick 3 visible red points and calculate:

- Point 1 distance to red center: _____
- Point 2 distance to red center: _____
- Point 3 distance to red center: _____
- Average squared distance: _____

Task 3: Track Variance Reduction

The chart shows total variance values at each step.



Is variance increasing or decreasing? _____

Task 4: Convergence Detection

Discovery Question:

How do you know the algorithm has finished? What stops changing?

Task 5: Discover the Rules

By watching the 6-step sequence, you can figure out the algorithm!

The algorithm repeats two rules:

Rule 1 (Assignment): Each point _____

Rule 2 (Update): Each center moves to _____

Task 6: What Gets Optimized?

The algorithm is trying to minimize something. What?

The objective is to minimize: _____

Key Insight:

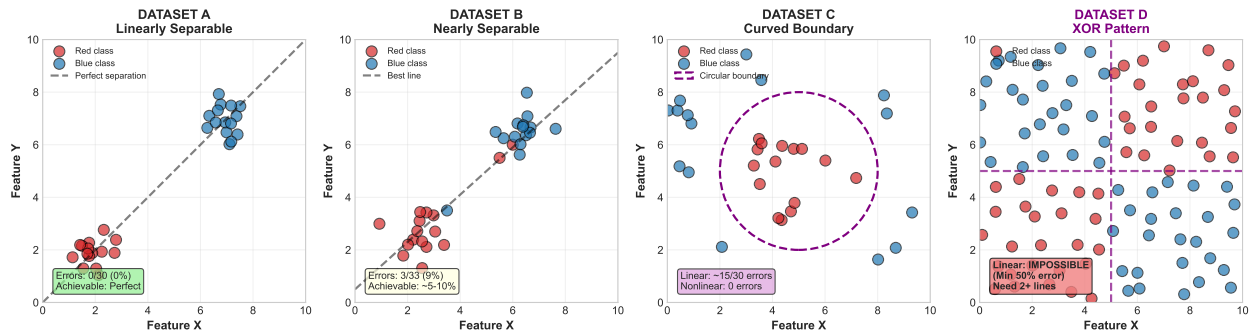
This “dance” has a name: **K-Means Clustering**

K = number of _____

Means = _____ position

It minimizes _____ within clusters

Discovery 3: The Impossible Separation



Observe the Chart

Four datasets show red and blue points. Your task: separate them with the simplest possible rule.

Discovery Tasks

Task 1: Draw Linear Boundaries

Using a ruler, draw the best straight line to separate red from blue in each dataset.

Count classification errors for each:

- Dataset A errors: ____/____ = ____%
- Dataset B errors: ____/____ = ____%
- Dataset C errors: ____/____ = ____%
- Dataset D errors: ____/____ = ____%

Which dataset(s) CAN be perfectly separated with a straight line? _____

Task 2: Mathematical Proof (Dataset D - XOR Pattern)

Any straight line has equation: $ax + by + c = 0$

Points are classified as:

- Red if $ax + by + c > 0$
- Blue if $ax + by + c < 0$

Test the four corners of Dataset D:

Point	Actual Color	$ax + by + c$	Predicted Color
(1,1)	Red	_____	_____
(1,9)	Blue	_____	_____
(9,1)	Blue	_____	_____
(9,9)	Red	_____	_____

Proof by Contradiction:

Assume (1,1) is Red: $a + b + c > 0$

Then (9,9) is Red: $9a + 9b + c > 0$

But (1,9) is Blue: $a + 9b + c < 0$

And (9,1) is Blue: $9a + b + c < 0$

Add last two inequalities: $(a + 9b + c) + (9a + b + c) < 0$

Simplifies to: $10a + 10b + 2c < 0$

But we need: $9a + 9b + c > 0$ (for point (9,9) to be Red)

Contradiction! No single line can separate XOR pattern.

Task 3: Nonlinear Solutions

For Dataset C (circular pattern), what shape boundary works?

Boundary type: _____

Equation: $x^2 + y^2 =$ _____

For Dataset D (XOR pattern), propose a solution using TWO lines:

Line 1: $x =$ _____ (vertical)

Line 2: $y =$ _____ (horizontal)

Combined rule: Point is Red if _____

Task 4: When Linearity Fails

Key Discoveries:

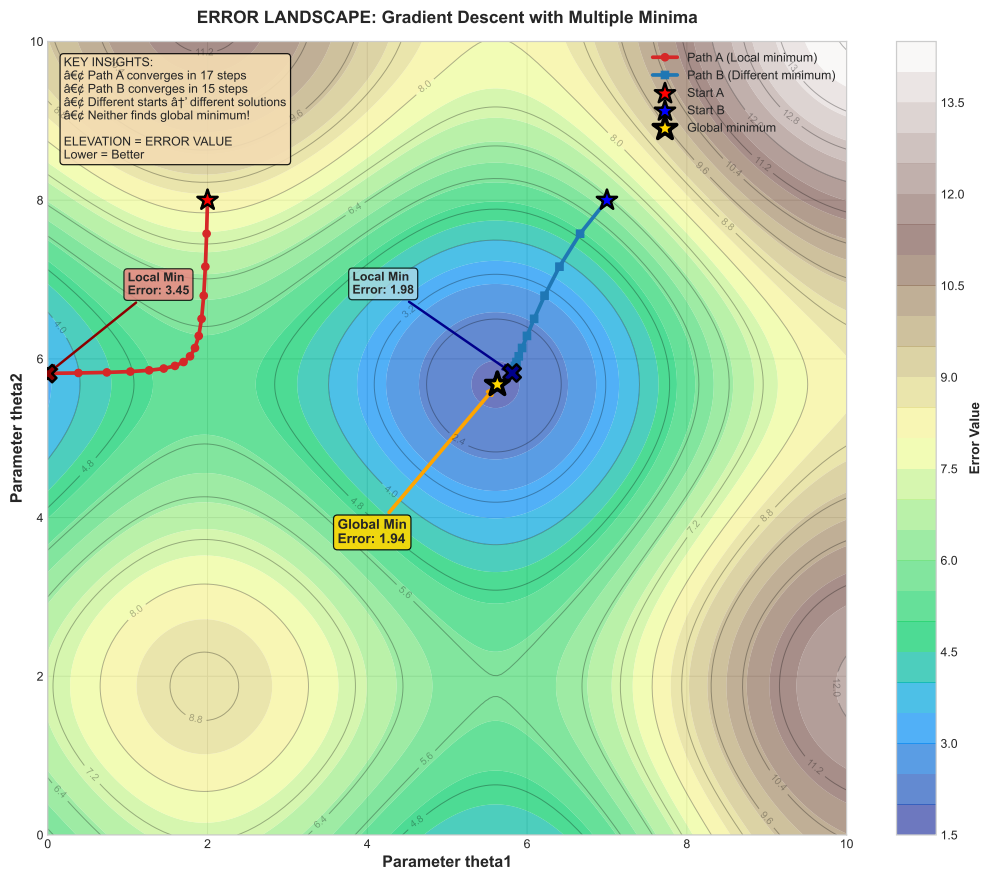
Linear models work when: _____

Linear models fail when: _____

Solution for complex patterns: _____

In the lecture, you'll learn how neural networks combine multiple lines to solve XOR

Discovery 4: The Optimization Landscape



Observe the Chart

This topographic map shows “error elevation.” Two paths (Red and Blue) descend from different starting points. Contour lines show equal error values.

Discovery Tasks

Task 1: Read the Terrain

- Path A (Red) starts at coordinates: (____, ____)
- Path A ends in valley with error: _____
- Path B (Blue) starts at coordinates: (____, ____)
- Path B ends in valley with error: _____

Did both paths find the SAME minimum? _____
Global minimum (gold star) has error: _____

Task 2: Calculate Gradients

Gradient = rate of change = slope

Pick a point on Path A around coordinates (3, 7):

Read contour values:

- Error at (3.0, 7): $E_1 \approx$ _____
- Error at (3.5, 7): $E_2 \approx$ _____
- Error at (3, 7.5): $E_3 \approx$ _____

Calculate gradients:

$$\frac{\partial E}{\partial x} \approx \frac{E_2 - E_1}{0.5} = \underline{\hspace{2cm}}$$

$$\frac{\partial E}{\partial y} \approx \frac{E_3 - E_1}{0.5} = \underline{\hspace{2cm}}$$

Descent direction (negative of gradient): $(-\frac{\partial E}{\partial x}, -\frac{\partial E}{\partial y}) = (\underline{\hspace{1cm}}, \underline{\hspace{1cm}})$

Task 3: Step Size Experiments

If learning rate (step size) is TOO LARGE:

- Problem: _____
- Risk: _____

If learning rate is TOO SMALL:

- Problem: _____
- Risk: _____

Optimal strategy: _____

Task 4: Local vs Global Minima

Discovery Question:

Why did Path A get stuck in a local minimum instead of finding the global minimum?

How could we help Path A escape its valley and find the better solution?

Task 5: Optimization Strategy

Update rule for gradient descent:

$$\theta_{\text{new}} = \theta_{\text{old}} - \alpha \cdot \nabla E(\theta)$$

where:

- θ = parameters (model weights) = _____
- α = learning rate (step size) = _____
- ∇E = gradient (direction) = _____

If gradient is positive, we move _____ (left/right)

If gradient is negative, we move _____ (left/right)

Key Insights:

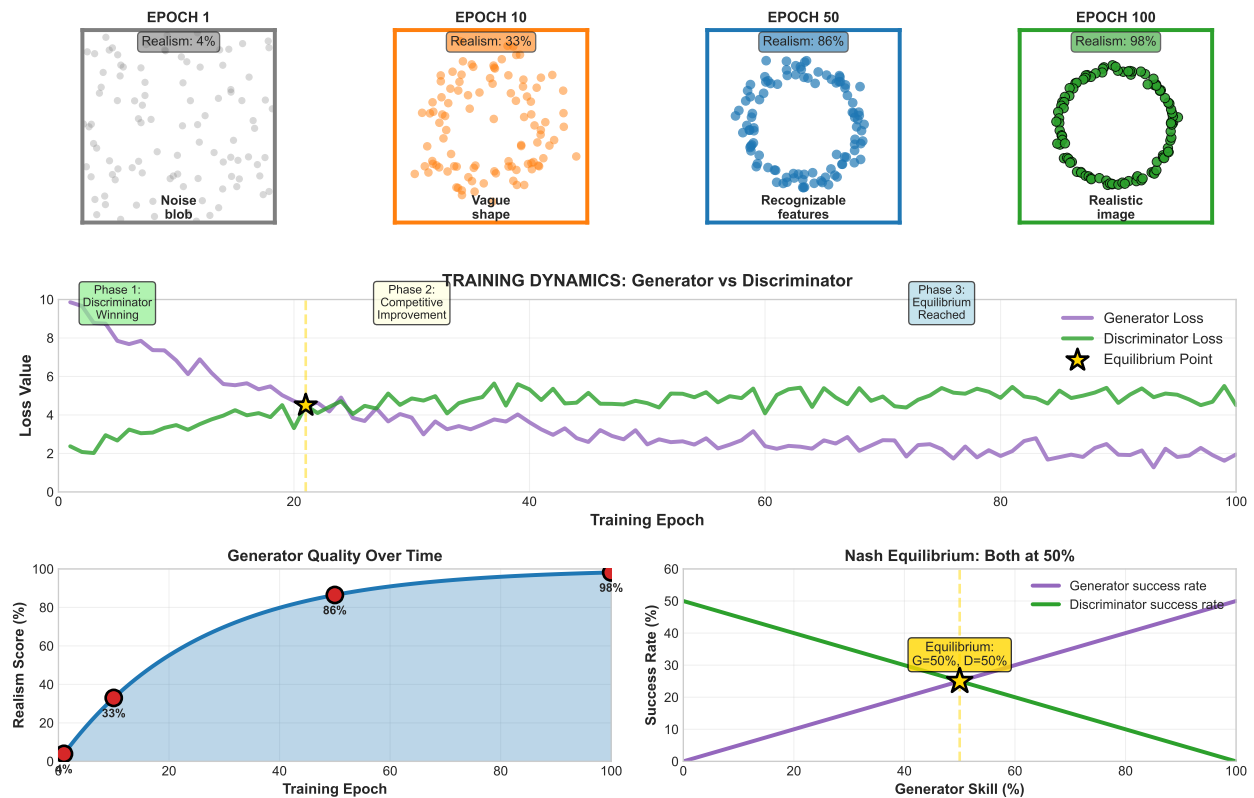
Gradient descent follows the $-\nabla L(\theta)$ direction

Different starting points lead to different solutions

Solution: Random restarts, momentum, or

Discovery 5: The Two-Player Game

GAN TRAINING EVOLUTION: The Two-Player Game



Observe the Chart

Top row shows generated samples improving from random noise to realistic. Middle shows loss curves for two competing players. Bottom shows quality metrics.

Discovery Tasks

Task 1: Track Quality Improvement

From the top row, read realism scores:

- Epoch 1 realism: _____% (looks like: _____)
- Epoch 10 realism: _____% (improvement: _____%)
- Epoch 50 realism: _____% (improvement: _____%)
- Epoch 100 realism: _____% (improvement: _____%)

Total improvement from Epoch 1 to 100: _____%

Task 2: Analyze Loss Dynamics

From the middle training curve:

When is the Generator (purple) winning? Epochs: _____

When is the Discriminator (green) winning? Epochs: _____

When do they reach equilibrium? Around epoch: _____

At equilibrium:

- Generator loss \approx _____
- Discriminator loss \approx _____

- Are they equal? _____

Task 3: Game Theory Calculations

From the bottom-right Nash equilibrium chart:

The formula for success rates:

- Generator success = $G \times (1 - D)$ where G = generator skill, D = discriminator skill
- Discriminator success = $D \times (1 - G)$

Fill in the table:

G skill	D skill	G success	D success	Who wins?
0.2	0.8	_____	_____	_____
0.5	0.5	_____	_____	_____
0.8	0.2	_____	_____	_____
0.9	0.1	_____	_____	_____

At what skill levels are both players equally successful? $G = \underline{\hspace{2cm}}$, $D = \underline{\hspace{2cm}}$

This is called **Nash Equilibrium**.

Task 4: Training Evolution

From the quality curve (bottom-left):

- Steepest improvement happens between epochs: _____ and _____
- Improvement slows down after epoch: _____
- Will realism ever reach 100%? _____ (Why/why not? _____)

Task 5: The Adversarial Insight

Discovery Questions:

Why do BOTH players improve even though only one can win each round?

If we trained only the Generator (no Discriminator feedback), what would happen?

The art student analogy: Generator = _____, Discriminator = _____

Key Insights:

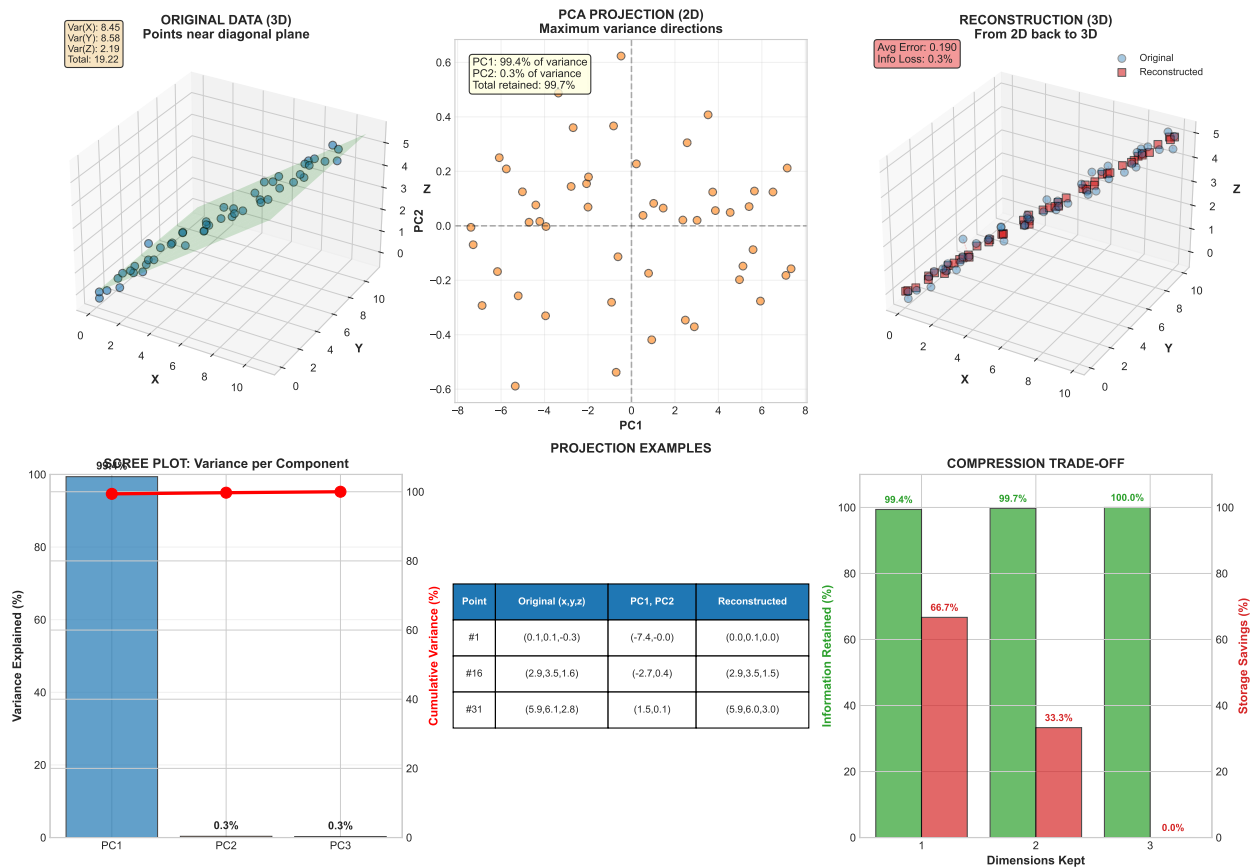
Competition drives _____

Equilibrium occurs at _____

This approach is called **Generative Adversarial Networks (GANs)**

Discovery 6: The Dimensionality Revelation

DIMENSIONALITY REDUCTION: From 3D to 2D with PCA



Observe the Chart

Top-left shows 3D data. Top-middle shows 2D projection. Top-right shows reconstruction error. Bottom shows variance accounting.

Discovery Tasks

Task 1: Variance Calculations

From the top-left 3D plot, read variance values:

- $\text{Var}(X) =$ _____
- $\text{Var}(Y) =$ _____
- $\text{Var}(Z) =$ _____
- Total variance = _____

From the top-middle 2D projection:

- $\text{Var}(\text{PC1}) =$ _____ (_____ % of total)
- $\text{Var}(\text{PC2}) =$ _____ (_____ % of total)
- Total retained = _____ %

Information lost by using only 2D: _____ %

Task 2: Reconstruction Error

From the table in bottom-middle, pick one sample point:

- Original 3D: (_____, _____, _____)
- Projected to 2D: (_____, _____)
- Reconstructed 3D: (_____, _____, _____)

Calculate error:

$$\text{Error} = \sqrt{(x_{orig} - x_{recon})^2 + (y_{orig} - y_{recon})^2 + (z_{orig} - z_{recon})^2} = \underline{\hspace{2cm}}$$

Average error from chart: _____

Is reconstruction good or bad? _____

Task 3: Compression Analysis

From bottom-right chart:

Dimensions	Info Retained	Storage Saved	Good Trade-off?
3 (original)	100%	0%	N/A
2	_____ %	_____ %	_____
1	_____ %	_____ %	_____

Calculate compression ratio for 2D:

Original storage: 50 points \times 3 dimensions = 150 numbers

Compressed storage: 50 points \times 2 dimensions = _____ numbers

Compression ratio: _____ (_____ % reduction)

With only _____ % information loss!

Task 4: When Does PCA Work?

Discovery Questions:

Why does this dataset compress so well from 3D to 2D?

Hint: Look at the 3D visualization - where do most points lie?

If points were randomly scattered in all 3 dimensions, would PCA work as well?

Task 5: Principal Component Direction

From the scree plot (bottom-left):

- PC1 captures: _____ % of variance (MOST important direction)
- PC2 captures: _____ % of variance (SECOND most important)
- PC3 captures: _____ % of variance (LEAST important)

Sum of all three: _____ % (should be 100%)

The “elbow” in the scree plot suggests keeping _____ components.

Key Insights:

PCA finds directions of _____ variance

Data lying near a lower-dimensional _____ compresses well

Trade-off: Storage savings vs _____

This technique is called **Principal Component Analysis (PCA)**

Final Reflection: Connecting Your Discoveries

Cross-Discovery Patterns

1. Optimization Appears Everywhere

You discovered optimization in multiple places:

- Discovery 1: Finding the model that balances training and test error
- Discovery 2: K-means minimizes within-cluster _____
- Discovery 4: Gradient descent minimizes _____
- Discovery 5: GANs optimize a competitive _____

What do all these have in common? _____

2. The Complexity Trade-off

- Discovery 1: Too simple (bias) vs too complex (_____)
- Discovery 3: Linear (simple) vs nonlinear (_____)
- Discovery 6: More dimensions (complex) vs fewer dimensions (_____)

General principle: _____

3. When Simple Rules Fail

- Discovery 3 Dataset D: Single line cannot solve _____
- Solution shown: Combine multiple _____
- This motivates: Neural networks with multiple _____

Prepare for Lecture

You've Discovered the Core Ideas!

In the lecture, you'll learn the formal names:

- Discovery 1 → **Bias-Variance Tradeoff**
- Discovery 2 → **K-Means Clustering**
- Discovery 3 → **Linear Separability & Neural Networks**
- Discovery 4 → **Gradient Descent Optimization**
- Discovery 5 → **Generative Adversarial Networks (GANs)**
- Discovery 6 → **Principal Component Analysis (PCA)**

The formulas will make sense because you've already discovered the patterns!

Three Most Important Insights

What are the three most surprising or important patterns you discovered?

1. _____

2. _____

3. _____

Questions for Lecture

What questions do you still have? Write 2-3 questions to ask during the lecture:

1. _____
2. _____
3. _____

Excellent work! Bring this completed worksheet to class.