

# Word Embeddings in 3D: Post-Class Learning Verification

Checking Your Understanding After the Interactive Lab

*From Words to Vectors: Can You Apply What You Learned?*

NLP Course 2025 - BSc Level Assessment

**Time Required:** 45-60 minutes

**Purpose:** Verify and deepen your understanding of word embeddings after completing the interactive notebook

**Format:** No coding required - focus on concepts, visualization, and application

## Checkpoint

**Before Starting:** You should have completed the “word\_embeddings\_3d\_bsc.ipynb” notebook. This handout will test your understanding of:

- Why words need to be vectors
- How Word2Vec learns from context
- Word similarity and clustering
- Word arithmetic
- Applications of embeddings

## Part A: Conceptual Understanding

(20 minutes)

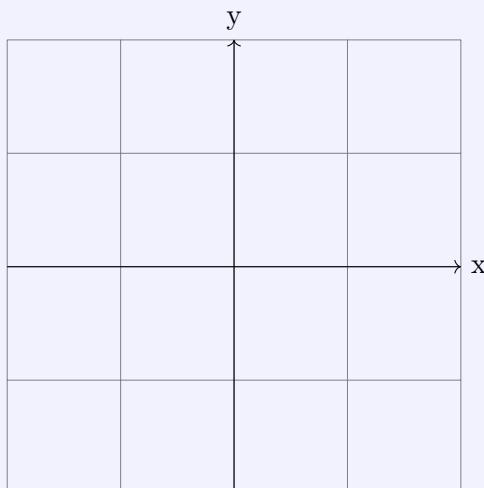
### A1: The Embedding Concept (5 minutes)

#### Why Vectors?

**Question 1:** Explain in your own words why computers need word embeddings instead of just treating words as text strings.

**Question 2:** Draw simple 2D vectors for these words showing their relationships:

- cat, dog, car
- Show which two should be closer together and why



*Draw vectors for cat, dog, and car*

**Question 3:** Circle the TRUE statements:

- Embeddings capture word meaning as numbers
- Similar words have similar vectors
- Each word gets exactly one dimension
- Context determines embedding values
- Embeddings are always 3D

## A2: Context Windows (5 minutes)

### Understanding Context

Given the sentence: “**The quick brown fox jumps over the lazy dog**”

**Task 1:** If we’re training on the word “fox” with window size = 2, circle all context words:

The quick brown **fox** jumps over the lazy dog

**Task 2:** How would changing window size affect learning?

Window = 1	_____
Window = 5	_____
Window = 10	_____

**Task 3:** Which window size would be better for:

- Learning syntax (grammar): \_\_\_\_\_
- Learning topic/theme: \_\_\_\_\_

## A3: Dimensions and Quality (5 minutes)

### Dimension Trade-offs

**Scenario:** You’re choosing embedding dimensions for different applications.

**Task 1:** Match the dimension size to the use case:

Application	Suggested Dimensions
Simple word similarity	3
Complex language model	10-50
Visualization in 3D	100-300
Mobile app (limited memory)	500+
Research with huge vocabulary	exactly 3

Options: 3, 10-50, 100-300, 500+, exactly 3

**Task 2:** Explain the trade-off:

- More dimensions = \_\_\_\_\_
- Fewer dimensions = \_\_\_\_\_

## A4: Quick Concept Check (5 minutes)

### Think About It

Rate your understanding (1 = confused, 5 = confident):

- |   |                     |
|---|---------------------|
| <input type="checkbox"/> Words as vectors       | [1] [2] [3] [4] [5] |
| <input type="checkbox"/> Context windows        | [1] [2] [3] [4] [5] |
| <input type="checkbox"/> Training process       | [1] [2] [3] [4] [5] |
| <input type="checkbox"/> Similarity measurement | [1] [2] [3] [4] [5] |

## Part B: Practical Application

(15 minutes)

### B1: Word Similarity Exercise (5 minutes)

#### Computing Similarity

Given these simplified 3D embeddings:

- king = [0.8, 0.2, 0.5]
- queen = [0.7, 0.3, 0.6]
- car = [0.1, 0.9, 0.2]

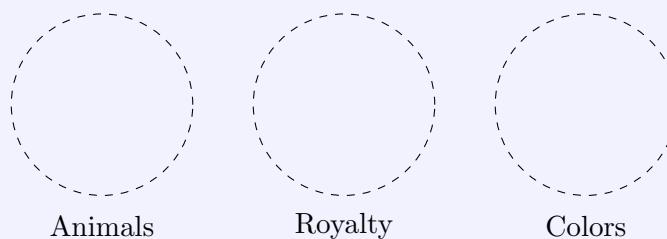
**Task 1:** Which pair is more similar? (Use rough estimation)

- king & queen: Distance  $\approx$  \_\_\_\_\_
- king & car: Distance  $\approx$  \_\_\_\_\_
- More similar pair: \_\_\_\_\_

**Task 2:** Rank these word pairs by expected similarity (1 = most similar):

- cat - dog
- king - queen
- happy - sad
- computer - laptop
- run - blue

**Task 3:** Draw approximate clusters:



*Place these words: cat, king, red, dog, queen, blue, prince, green*

## B2: Word Arithmetic Magic (5 minutes)

### Vector Math with Words

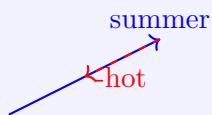
**Task 1:** Complete these analogies:

- king - man + woman = \_\_\_\_\_
- paris - france + germany = \_\_\_\_\_
- cat - kitten + puppy = \_\_\_\_\_

**Task 2:** Create your own word equation:

\_\_\_\_\_ - \_\_\_\_\_ + \_\_\_\_\_ = \_\_\_\_\_

**Task 3:** Draw the vector arithmetic for: “summer - hot + cold”



*Complete the diagram*

**Task 4:** Explain why word arithmetic works:

## B3: 3D Visualization Interpretation (5 minutes)

### Reading 3D Plots

Imagine you see a 3D plot where:

- “love” and “hate” are far apart
- “cat” and “dog” are close together
- “king” and “queen” form a cluster with “prince”

**Task 1:** What does distance represent in the plot?

**Task 2:** Where would you expect to find these words?

- “affection” - Near: \_\_\_\_\_
- “princess” - Near: \_\_\_\_\_
- “fish” - Near: \_\_\_\_\_

**Task 3:** If words gradually move closer during training, what’s happening?

## Part C: Hands-On Problem Solving

(15 minutes)

### C1: Build Your Own Embeddings (5 minutes)

#### Design Embeddings from Scratch

Given these 5 sentences:

1. The cat sleeps
2. The dog plays
3. Cats and dogs play
4. Birds fly high
5. Fish swim deep

**Task 1:** Create word-context pairs for “cat” (window=1):

- Context words: \_\_\_\_\_
- Context words: \_\_\_\_\_

**Task 2:** Design simple 2D vectors for these words:

Word	x	y
cat		
dog		
bird		
fish		

**Task 3:** Which words should cluster together? Why?

## C2: Application Design (5 minutes)

### Building with Embeddings

**Task 1:** Design a synonym finder:

1. Input: \_\_\_\_\_
2. Process: \_\_\_\_\_
3. Output: \_\_\_\_\_

**Task 2:** Sketch a sentiment analyzer:

How would you use embeddings to determine if text is positive/negative?

**Task 3:** Your creative application:

Design a new use for word embeddings:

- Name: \_\_\_\_\_
- Purpose: \_\_\_\_\_
- How embeddings help: \_\_\_\_\_

## C3: Debugging Scenarios (5 minutes)

### Common Pitfall

Real problems you might encounter:

### Problem Solving

**Scenario 1:** The word “bank” appears near both “river” and “money”.

- Problem: \_\_\_\_\_
- Solution: \_\_\_\_\_

**Scenario 2:** A new word “COVID” doesn’t exist in your embeddings.

- Problem: \_\_\_\_\_
- Solution: \_\_\_\_\_

**Scenario 3:** Your embeddings show “doctor”=male, “nurse”=female bias.

- Problem: \_\_\_\_\_
- Solution: \_\_\_\_\_

## Part D: Reflection & Extension

(10 minutes)

### D1: Self-Assessment Checklist (3 minutes)

#### Checkpoint

Check off what you can now do:

- Explain why words need to be vectors
- Describe how Word2Vec learns from context
- Calculate word similarity (roughly)
- Perform word arithmetic
- Identify word clusters in 3D space
- Design applications using embeddings
- Recognize common problems and solutions

### D2: Real-World Connections (3 minutes)

#### Real World Application

##### Connect to Industry:

1. How does Google use embeddings in search?

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2. How do embeddings connect to ChatGPT/BERT?

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3. Name one ethical concern with embeddings:

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### D3: Challenge Questions (4 minutes)

#### Think About It

##### Going Deeper:

1. Could we create embeddings for images? How?
2. What about embedding DNA sequences or music?
3. How would you embed an entire document (not just words)?

### Final Reflection

#### Discovery Moment

##### The Big Picture:

You've learned that word embeddings transform language into mathematical space where:

- Meaning becomes measurable
- Relationships become computable
- Patterns become visible

This is the foundation of modern NLP - from search engines to chatbots to translation systems!

**Areas needing review?** List them here:

- \_\_\_\_\_
- \_\_\_\_\_
- \_\_\_\_\_

**Most interesting discovery:** \_\_\_\_\_

**One question you still have:** \_\_\_\_\_

## Next Steps

- Review sections with ratings below 3
- Try coding your own Word2Vec model
- Explore pre-trained embeddings (GloVe, FastText)
- Learn about contextual embeddings (BERT)
- Apply embeddings to a real project

— End of Assessment —

*Instructor: Check answer key for grading rubric and common misconceptions*