

Homework 3: Text Processing Fundamentals

Points: 20 | **Due:** Sunday, February 15, 2026 @ 11pm Pacific

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Compute: CPU (free tier)

Learning Objectives

1. **Install and use** NLP libraries (spaCy, NLTK) in Google Colab
 2. **Understand** WHY we preprocess text (not just how)
 3. **Create** domain-specific stopwords for your data
 4. **Measure** the impact of text cleaning on analysis
 5. **Identify** cases where cleaning hurts your analysis
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Grading

| Component | Points | What We're Looking For |
|--------------------|-----------|-------------------------------------|
| Environment Setup | 3 | NLP libraries installed and working |
| Standard Stopwords | 4 | Applied and analyzed impact |
| Domain Stopwords | 5 | Created 10+ with justification |
| Negation Analysis | 5 | Smart stopwords preserving meaning |
| Visualization | 3 | Word cloud comparison |
| Total | 20 | |

The Big Picture

Text preprocessing is like preparing ingredients before cooking. But just like cooking, **the same preparation isn't right for every dish.**

Standard stopword removal can destroy important meaning (like negations in sentiment analysis).

Instructions

1. Open MIS769_HW3_Text_Processing.ipynb in Google Colab
 2. Load your dataset (HuggingFace, Kaggle, or your own)
 3. Apply standard stopword removal and analyze what was removed
 4. Create your own domain-specific stopwords with justification
 5. Analyze the negation problem and create a “smart” stopword list
 6. Generate word cloud visualizations comparing approaches
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Questions to Answer

- **Q1:** Which removed stopwords might carry meaning in your domain?
 - **Q2:** Why did you choose each domain-specific stopword?
 - **Q3:** When should you preserve vs. remove negations?
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Submission

Upload to Canvas: - Your completed .ipynb notebook with all cells executed

— *Richard Young, Ph.D.*