

# Welcome to NLP 2025

From N-grams to Transformers: Your Journey Begins Today

NLP Course 2025

First Day of Class

Natural Language Processing: 12 Weeks to Understanding ChatGPT

# What You'll Build This Semester

## From Zero to Transformer

### Week 2:

- Word embeddings from scratch
- Train on real text
- Visualize word similarities

### Week 3:

- LSTM language model
- Generate Shakespeare
- Understanding memory

### Week 5:

- **Complete transformer**
- Self-attention mechanism
- The architecture behind ChatGPT

### Week 6:

- Fine-tune BERT
- Transfer learning
- State-of-the-art results

### Week 9:

- Text generation strategies
- Control creativity
- Beam search and sampling

### Week 12:

- Deploy responsibly
- Measure and mitigate bias
- **Ethical AI**

# Why This Course Matters

## NLP is Transforming Every Industry

### Current Applications:

- **ChatGPT & Claude** - Conversational AI
- **GitHub Copilot** - Code generation
- **Google Translate** - 100+ languages
- **Grammarly** - Writing assistance
- **Alexa & Siri** - Voice assistants
- **Gmail** - Smart compose

### Career Impact:

- NLP engineers: Top 5 percent salaries
- ML researchers: High demand
- Data scientists: Essential skill
- Product managers: Understanding capability
- All roles: AI literacy critical

### Market Growth:

- NLP market: USD 20B (2024)
- Projected: USD 100B+ (2030)
- AI investment: USD 500B+ annually

Understanding transformers is understanding the future of computing

# Our Learning Philosophy

## Learn by Building, Not Just Watching

### Discovery-Based Learning:

- Start with concrete problems
- Discover solutions yourself
- THEN learn the formal theory
- Build intuition first

### Hands-On Practice:

- 12 lab notebooks
- Real code, real data
- Run in your browser
- No magic black boxes

### Progressive Complexity:

- Week 1: Count words
- Week 5: Build transformer
- Week 12: Deploy ethically
- Each week builds on previous

### Real Implementation:

- PyTorch from scratch
- Every line explained
- No hidden abstractions
- Understanding not memorization

**By the end, you'll have IMPLEMENTED a transformer, not just understood it**

# Your 12-Week Journey

## Three Phases of Mastery

### Phase 1: Foundations

Weeks 1-4

**Week 1:** Statistical LM

N-grams, probability

**Week 2:** Word Embeddings

Word2Vec, GloVe

**Week 3:** RNN/LSTM

Sequential models

**Week 4:** Seq2Seq

Attention mechanism

Master: Neural basics

### Phase 2: Revolution

Weeks 5-8

**Week 5:** Transformers

Self-attention

**Week 6:** Pre-trained

BERT, GPT

**Week 7:** Advanced

T5, GPT-3, scaling

**Week 8:** Tokenization

BPE, WordPiece

Master: Modern architectures

### Phase 3: Application

Weeks 9-12

**Week 9:** Decoding

Generation strategies

**Week 10:** Fine-tuning

Adaptation methods

**Week 11:** Efficiency

Optimization

**Week 12:** Ethics

Responsible deployment

Master: Production deployment

Each phase unlocks new capabilities - by Week 12, you're deployment-ready

# Prerequisites: Do You Have What It Takes?

## Self-Assessment

### Required (Must Have):

- Python programming experience
- Basic linear algebra (vectors, matrices)
- Basic probability theory
- Comfortable with loops and functions
- Can read and write code

### If You Don't Have These:

- Python: Complete Python tutorial first
- Math: Review Khan Academy linear algebra
- Probability: Review basic probability

### Helpful But Not Required:

- PyTorch or TensorFlow
- Neural networks basics
- Machine learning concepts
- Backpropagation understanding
- GPU programming

### We Provide:

- Neural Network Primer (if needed)
- LSTM Primer (deep dive)
- Progressive difficulty
- Office hours support

**New to neural networks? Start with our Neural Network Primer before Week 2**

# Course Materials: Everything You Need

## Complete Open-Source Course

### Presentations:

- 60+ slide decks
- Optimal readability design
- Multiple versions (BSc, enhanced)
- PDF and LaTeX source

### Lab Notebooks:

- 12 interactive Jupyter notebooks
- Hands-on implementation
- Real data and models
- Run in your browser

### Handouts:

- Pre-class discovery exercises
- Post-class technical practice
- Student and instructor versions

### Supplementary:

- Neural Network Primer
- LSTM Primer (32 slides)
- Word Embeddings Module
- 8 visualization notebooks

### Documentation:

- Complete installation guide
- Week-by-week course index
- Troubleshooting support
- Project templates

### All Materials:

- Open source (MIT license)
- GitHub repository
- Always accessible

# Getting Started TODAY

## Your First Assignment (Due Next Class)

### Step 1: Clone Repository

- `git clone`  
`github.com/josterri/2025_NLP_Lectures.git`
- `cd 2025_NLP_Lectures`

### Step 2: Install Dependencies

- `pip install -r requirements.txt`

### Step 3: Verify Installation

- `python verify_installation.py`

### Expected Time:

- Clone: 2 minutes
- Install: 10-15 minutes
- Verify: 30 seconds

### Step 4: Test First Notebook

- Launch: `jupyter lab`
- Open: Week 2 word embeddings lab
- Run first 3 cells
- Verify imports work

### GPU Optional:

- Weeks 1-4: CPU sufficient
- Weeks 5+: GPU recommended
- All labs work on CPU
- GPU speeds up training

### Get Help:

- Read `INSTALLATION.md`
- Check GitHub Issues
- Office hours: TBD



# Week-by-Week Schedule

## What Happens Each Week

### Before Class (Monday):

- Read handout (pre-class section)
- Complete discovery exercises
- Build intuition
- Identify questions

### During Lecture (Wednesday):

- Review key concepts
- Live coding demonstrations
- Q&A on pre-class material
- Preview lab notebook

### Lab Session (Friday):

- Implement concepts
- Hands-on coding
- Get help from TAs

### After Lab (Weekend):

- Complete lab notebook
- Finish post-class handout
- Experiment and explore
- Submit by Monday

### Office Hours (TBD):

- Individual help
- Concept clarification
- Project discussion
- Career advice

### Weekly Commitment:

- Lecture: 2 hours
- Lab: 2 hours
- Independent work: 4-6 hours

# How You'll Be Assessed

## Grading Breakdown

### Lab Notebooks (40 percent):

- 12 notebooks, 11 graded (drop lowest)
- Due Monday after lab
- Implementation correctness
- Code quality and comments
- Experimentation and insights

### Midterm Project (25 percent):

- After Week 6
- Implement and evaluate model
- Written report
- Code submission

### Final Project (30 percent):

- Weeks 11-12
- Original NLP application

### Participation (5 percent):

- Class attendance
- Lab participation
- Office hours engagement
- Helping classmates

### Grading Philosophy:

- Focus on learning not perfection
- Partial credit for attempts
- Bonus for creativity
- Collaboration encouraged (cite sources)

### Late Policy:

- 48-hour grace period (no penalty)
- After that: 10 percent per day
- Max 5 days late

# Example Final Projects

## What Past Students Built

### Text Generation:

- **Poetry generator** - Style transfer
- **Code completion** - GitHub-trained
- **Story writer** - Character-consistent
- **Email composer** - Professional tone

### Information Extraction:

- **Resume parser** - Extract skills
- **News summarizer** - Multi-document
- **Question answering** - Domain-specific
- **Fact checker** - Claim verification

### Classification:

- **Sentiment analysis** - Product reviews
- **Spam detection** - Email filtering
- **Intent classification** - Chatbot

### Translation & Multilingual:

- **Domain translator** - Legal/medical
- **Code-switching** - Multilingual text
- **Style transfer** - Formal to casual
- **Dialect conversion** - Regional

### Creative Applications:

- **Music from text** - Lyrics to melody
- **Debate partner** - Argument generation
- **Language learning** - Adaptive tutor
- **Game dialogue** - NPC conversations

### Research Projects:

- **Bias measurement** - Gender/race
- **Interpretability** - Attention analysis
- **Efficiency** - Model compression

# Milestone: What You'll Know By Week 5

## From Zero to Transformer in 5 Weeks

### Technical Understanding:

- How attention mechanism works
- Query, Key, Value matrices
- Multi-head attention computation
- Positional encoding mathematics
- Layer normalization
- Feed-forward networks
- Residual connections

### Implementation Skills:

- Scaled dot-product attention
- Multi-head splitting and merging
- Complete transformer block
- Positional encoding
- Training loop

### Practical Abilities:

- Debug attention weights
- Visualize attention patterns
- Optimize hyperparameters
- Compare with RNN/LSTM
- Understand GPT architecture
- Read research papers

### Real-World Knowledge:

- Why ChatGPT is so powerful
- How BERT differs from GPT
- Computational complexity trade-offs
- When to use transformers
- Current limitations
- Future directions

# Common Questions

## Frequently Asked Questions

### **Q: Do I need a GPU?**

A: No, all labs work on CPU. GPU speeds things up for Weeks 5+ but isn't required.

### **Q: How much Python?**

A: Comfortable writing functions, loops, and classes. PyTorch will be taught.

### **Q: Mathematical background?**

A: Linear algebra (vectors, matrices, dot products) and basic probability. We'll review as needed.

### **Q: Can I audit?**

A: Yes! All materials are open source. You won't get grades but can follow along.

### **Q: Group projects?**

A: Labs are individual. Final project can be pairs with instructor approval.

### **Q: How hard is this?**

A: Challenging but doable. Budget 10 hours/week. We support you every step.

### **Q: What if I fall behind?**

A: Office hours, catch-up sessions, and grace periods. Communication is key.

### **Q: Industry vs research?**

A: Both! Course covers practical deployment and research methods.

### **Q: After this course?**

A: You'll be ready for NLP engineering roles, ML research, or advanced courses.

### **Q: Most important week?**

A: Week 5 (transformers) is the foundation. Don't miss it!

# Strategies for Success

## How Top Students Excel

### Before Each Week:

- Read pre-class handout carefully
- Try exercises before looking at solutions
- Write down questions
- Review previous week's concepts

### During Lab:

- Start early (don't wait until deadline)
- Experiment beyond requirements
- Ask questions immediately
- Help classmates (best way to learn)
- Save your work frequently

### Study Groups:

- Form groups of 3-4 students
- Meet weekly to discuss concepts

### Debugging Mindset:

- Read error messages carefully
- Print intermediate outputs
- Test with small examples first
- Check dimensions and shapes
- Use debugger or print statements

### Going Deeper:

- Read suggested papers
- Try bonus challenges
- Implement variants
- Share findings with class

### Stay Organized:

- Keep notes for each week
- Maintain code repository

# Resources and Support

## Where to Get Help

### Course Materials:

- **GitHub:** [github.com/josterri/2025\\_NLP\\_Lectures](https://github.com/josterri/2025_NLP_Lectures)
- **Documentation:** README.md, COURSE\_INDEX.md
- **Installation:** INSTALLATION.md
- **Syllabus:** SYLLABUS.md

### Getting Help:

- **Office Hours:** TBD
- **Email:** TBD
- **Discussion Forum:** TBD
- **Lab TAs:** Available during lab sessions

### External Resources:

- **PyTorch Tutorials:** [pytorch.org/tutorials](https://pytorch.org/tutorials)
- **Papers:** [arxiv.org](https://arxiv.org)
- **Blogs:** Jay Alammam, Lil'Log

### Community:

- **GitHub Issues:** Report bugs
- **Pull Requests:** Contribute improvements
- **Discussions:** Share insights

### Reference Books:

- Speech and Language Processing (Jurafsky & Martin)
- Neural Network Methods for NLP (Goldberg)
- Dive into Deep Learning (Zhang et al)
- All available free online

### Tools:

- **Google Colab:** Free GPU notebooks
- **Hugging Face:** Pre-trained models
- **Weights & Biases:** Experiment tracking

# Your Immediate Action Items

## What to Do Right Now

### Today (Next 30 Minutes):

1. Clone the repository
2. Read the README.md
3. Check system requirements
4. Identify if you need GPU

### Tonight (1-2 Hours):

1. Install dependencies (requirements.txt)
2. Run verify\_installation.py
3. Fix any installation issues
4. Read INSTALLATION.md if stuck

### Before Next Class:

1. Complete Neural Network Primer (if needed)
2. Read Week 2 pre-class handout
3. Try the discovery exercises
4. Prepare questions

### Optional (For Eager Students):

1. Browse Week 2 lab notebook
2. Read Week 1 presentation
3. Check out supplementary modules
4. Join discussion forum

**Most important: Get your environment working BEFORE Week 2**



# Understanding the AI Revolution

## Why Transformers Changed Everything

### Before Transformers (Pre-2017):

- Sequential processing (RNNs)
- Slow training (hours to days)
- Limited context (hundreds of words)
- Vanishing gradient problems
- Hard to parallelize
- Specialized architectures per task

### Limitations We Hit:

- Couldn't scale to large models
- Couldn't use massive datasets
- Couldn't capture long-range dependencies
- Transfer learning was limited

### After Transformers (2017+):

- Parallel processing (attention)
- Fast training (hours on GPUs)
- Unlimited context (thousands of tokens)
- Stable gradients
- Massively parallelizable
- Universal architecture

### What Became Possible:

- GPT-3: 175 billion parameters
- GPT-4: Multimodal understanding
- Claude: 100k+ token context
- ChatGPT: Conversational AI
- All modern LLMs use transformers

# What You'll Be Able to Do

## By the End of This Course

### Technical Skills:

- **Implement** transformers from scratch
- **Fine-tune** BERT and GPT models
- **Debug** attention mechanisms
- **Optimize** model efficiency
- **Deploy** models responsibly
- **Measure** and mitigate bias

### Conceptual Understanding:

- How ChatGPT works internally
- Why attention beats RNNs
- When to use which architecture
- Trade-offs in model design
- Current research frontiers

### Practical Abilities:

- Read and implement research papers
- Design NLP systems
- Evaluate model performance
- Choose appropriate methods
- Debug complex models
- Communicate technical concepts

### Career Readiness:

- NLP engineer interviews
- ML research positions
- Data science roles
- PhD program preparation
- Startup technical co-founder
- Technical leadership

# Next Week Preview: Word Embeddings

## Your First Hands-On Lab

### What You'll Learn:

- Words as vectors
- Semantic similarity
- Word2Vec algorithm (CBOW & Skip-gram)
- Training on real text
- Visualizing embeddings
- Analogies: king - man + woman = queen

### What You'll Build:

- Train word embeddings from scratch
- Implement Skip-gram model
- Visualize word relationships
- Discover semantic patterns
- Test word analogies

### Prepare By:

- Reading pre-class handout
- Understanding distributional hypothesis
- Reviewing basic neural networks
- (Optional) Neural Network Primer

### Lab Highlights:

- Work with real text corpus
- See embeddings evolve during training
- Interactive visualizations
- Explore semantic relationships
- Bonus challenges available

Your Journey Starts Now

# Welcome to NLP 2025

## Key Takeaways:

- **Build transformers** - Understand how ChatGPT works
- **12-week journey** - From foundations to deployment
- **Hands-on learning** - Implement everything yourself
- **Real applications** - Deploy models responsibly
- **Complete support** - Extensive materials and help

## First Assignment:

Install environment by next class  
Run `verify_installation.py` successfully  
Read Week 2 pre-class handout