

# DS 6050: Deep Learning Course Project Plan

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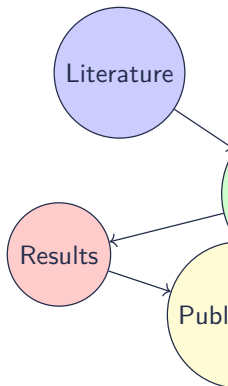
## Course Project: Grand Challenge

### Objective:

- Apply deep learning concepts to solve real-world problems
- Conduct rigorous research with ablation studies
- Produce publication-quality work

### Team Size:

- 2-3 members per team



## Critical: Dataset Requirements

### Public Dataset Requirement

**! Your dataset MUST be publicly available**

#### Why Public Datasets?

- Enable paper publication
- Allow result reproduction
- Foster academic transparency
- Build your portfolio

#### Not acceptable

- ND -pro
- Proprietary
- Private or
- Restricted

### Recommendation

Choose well-documented benchmark datasets or create your own dataset with proper documentation.

## Recommended Dataset Sources

### Computer Vision:

- ImageNet, COCO, CIFAR
- Medical: NIH Chest X-ray, ISIC
- Kaggle Vision Competitions
- Hugging Face Datasets

### NLP & Text:

- GLUE, SuperGLUE
- Common Crawl
- WikiText, BookCorpus
- arXiv papers dataset

### Time Series & Audio

- UCI ML Repository
- Google Speech
- LibriSpeech, LibriVox
- Financial data (e.g., Yahoo Finance)

### Multimodal:

- MS-COCO Caption
- Visual Question Answering
- CLIP datasets
- Video datasets (e.g., Kinetics)

**Pro Tip:** Include dataset URL in your proposal

# Project Timeline & Milestones



Milestone	Due (Module)
I: Literature Review & Proposal	Module 8
II: Architecture & Preliminary Results	Module 10
III: Final Report & Presentation	After Module 12

## Milestone I: Literature Review & Proposal

**Due: End of Module 8 (Attention & Seq2Seq)**

### Deliverables:

- 1 **Motivation:** Problem statement
- 2 **Dataset:** Public URL required
- 3 **Literature Review:**
  - Minimum 3 papers
  - Prior methodologies
  - Identify gaps
- 4 **Proposed Method:** Initial approach
- 5 **Experiments:** Evaluation plan

### Format Requirements:

- Use arXiv template
- 2-3 pages (excluding references)
- Proper citation style
- Clear research objectives

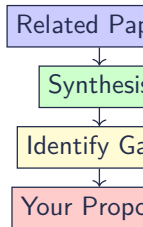
### Submission:

- Format: PDF
- Filename: GroupID\_proposal.pdf
- Submit via Canvas LMS

## Literature Review Best Practices

### What Makes a Good Literature Review?

- **Organized:** Directly related to your research question
- **Synthesized:** Summary of known vs unknown
- **Critical:** Identify controversies and gaps
- **Forward-looking:** Formulate new research questions



### Resources

- Google Scholar, arXiv, Papers with Code
- Conference proceedings: NeurIPS, ICML, CVPR, CL
- Focus on recent work (2020-2025)



## Milestone II: Model Architecture & Experimentation

Due: End of Module 10 (Vision Transformers)

### Key Components

- 1 **Abstract:** Concise project summary
- 2 **Introduction:** Problem context and motivation
- 3 **Literature Survey:** Expanded from Milestone I
- 4 **Method:** Detailed architecture description
- 5 **Preliminary Experiments:** Initial results
- 6 **Next Steps:** Planned improvements
- 7 **Member Contributions:** Individual responsibilities

### Required Elements:

- At least one baseline model
- Training curves & metrics
- Error analysis
- Ablation study plan

### Submission Format:

- Compact, up to 10 pages
- Include code repository link
- Jupyter notebook for experiments
- GroupID\_checklist

# Experimental Design Guidelines

## Baseline Selection:

- Start simple (e.g., ResNet for vision)
- Use pre-trained when applicable
- Document hyperparameters
- Ensure reproducibility

## ablation Studies:

- Vary one component at a time
- Document all changes
- Use consistent seeds
- Track with W&B or TensorBoard

## Metrics to Report:

- Task-specific metrics
- Training/validation loss
- Computational cost
- Parameter count
- Inference time

# Final Project Requirements

**Due:** After Module 12 (Generative Models)

## Deliverables:

### 1 Final Report (6-8 pages)

- Full methodology
- Comprehensive results
- Thorough analysis
- Future work

### 2 Code Repository

- Clean, documented code
- RE DME with instructions
- Requirements file

### 3 Presentation (5 minutes)

- Key findings
- Live demo if applicable
- Q&A session

## Evaluation Criteria:

- Technical depth
- Innovation
- Experimentation
- Result interpretation
- Presentation quality
- Reproducibility

## Peer Review:

- Review 2 other projects
- Provide constructive feedback
- Vote for "Best Project"

## Project Ideas by Module Coverage

### Computer Vision (Modules 4-5, 10):

- Medical image diagnosis
- Object detection improvements
- ViT vs CNN comparison
- Domain adaptation

### NLP (Modules 6-9):

- Text summarization
- Question answering
- Code generation
- Sentiment analysis

### Multimodal (Module 11):

- Image captioning
- Visual question answering
- CLIP-style training
- Video understanding

### Generative (Module 12):

- VAE for data generation
- GAN improvements
- Diffusion models
- Style transfer

### Innovation Opportunities

- Combine architectures. Reproduce advanced SoT
- Apply models to new domains
- Improve efficiency/speed
- Address ethical considerations

## Best Practices for Success

### Team Collaboration:

- Use Overleaf for LaTeX
- GitHub for version control
- Regular team meetings
- Clear task division

### Experiment Tracking:

- Use W&B or TensorBoard
- Document all hyperparameters
- Save model checkpoints
- Keep experiment logs

### Code Quality:

- Modular design
- Clear documentation
- Unit tests for key components
- Requirements.txt

### Time Management

- Start literature review early
- Allocate time for unexpected delays
- Plan for computation time
- Buffer for writing the paper

### Common Pitfalls to Avoid

- Choosing overly complex problems
- Ignoring baseline comparisons
- Poor reproducibility documentation
- Last-minute dataset changes

## Resources & Support

### Technical Resources:

- UV Rivanna HPC cluster
- Google Colab (free GPU)
- PyTorch documentation
- D2L textbook examples

### Office Hours:

- Dedicated project hours
- T support sessions
- Ed Discussion forum
- Peer collaboration

### Writing Resources:

- Overleaf template
- arXiv paper format
- Citation manager
- Academic writing guide

### Mentorship:

- Primary: Instructor
- Based on project needs
- Regular check-ins
- Feedback on milestones

**Note:** This project is your opportunity to create portfolio work!

## Questions & Discussion

**Go to the forum!**

### **Next Steps:**

- 1 Form teams (2-4) members
- 2 Explore different set options
- 3 Start literature review

## Key Takeaways

**Public Dataset**

! No ND da

**Follow Timeline**

3 milestones

**Research Quality**

blations req

**Reproducibility**

Clean code

**Good luck with your projects**