MSML640 Computer Vision

Project Progress Report

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GitHub Repository Link

https://github.com/willloe/MSML640_Group10

Summary of progress

For the classifier pipeline, text extraction from the SlideVQA dataset was expanded to include all text-related elements such as text boxes, titles, and page numbers, and an OCR system is being integrated to capture text with bounding box positions. A slide-level classifier was developed to detect whether slides contain diagrams, figures, or images, achieving an accuracy of 0.8. Future work is now focused on extending this to detect multiple elements per slide with bounding boxes and multi-label outputs. Metadata for the SlideVQA dataset was generated, and scripts were created to parse the dataset, extract slide images, and annotate them with bounding boxes. A web server is also being implemented to automate the process of parsing presentations into training data.

For the diffusion model, the project has a working SDXL-based generation stack with layout awareness and reproducible smoke tests. A preprocessing module converts slide layouts into a 4-channel control tensor and a safe-zone mask, enabling ControlNet-style guidance that respects title/body/image/logo regions while preserving background freedom. Compact JSON Schemas for layouts and palettes are in place, plus a synthetic data module that creates realistic, schema-valid samples for testing without external datasets. The inference pipeline loads SDXL with optional LoRA, applies a readability-focused safe-zone postprocess, and optionally uses ControlNet with a configurable control_from signal (element/safe/edge). Debug artifacts include the exact conditioning image, raw generation, masked output, overlay, and a determinism hash. Smoke scripts run locally (CPU skips heavy gen) and on Colab (T4 GPU), giving a clean, testable foundation for upcoming work on schedulers, ControlNet tuning, and LoRA fine-tuning.

New task division:

- William Loe (Generation stack): ControlNet conditioning MVP, scheduler options (DDIM/DPM++), LoRA training scaffold, 512 -> 1280 upscaler, and optional inpainting hook.
- HengMeng (Data/metrics/integration): WCAG-AA readability metric, layout-safety metric, dataset + manifest, synthetic recipe/themes, CLI + Colab UX, and reporting.

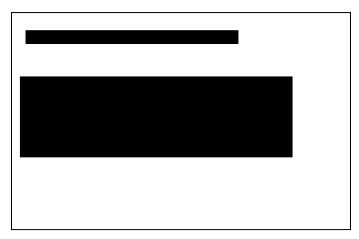
Proof-of-Concept or early results

Classifier Model:

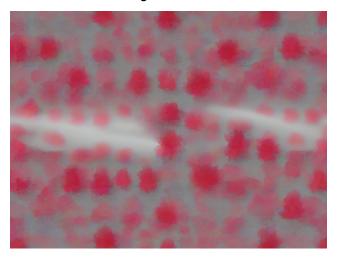
```
device: cuda
Classes: ['Diagram', 'Figure', 'Image']
Train Samples: 37741 Validate Samples: 9435
Epoch 01 | train_loss = 0.0002, train_accuracy = 0.789, validate_accuracy = 0.797
Training complete. Best validate accuracy is: 0.797
```

Diffusion Model:

Generated safe zone for text areas



Generated raw background



Early result: raw background + safe zone mask



Updates (if applicable)

Classifier Model:

Anirud: The extraction dataset code has been modified to extract all text-related content from the SlideVQA dataset, including TextBoxes, Titles, and Page Numbers. Additionally, an OCR extraction program will be implemented to extract text content along with its corresponding bounding box position on each slide.



AGENDA

- CEF at a glance
- CEF reuse logic
- CEF building blocks

Informatics

Hsing-Hao: Created a classifier using a dataset from SlideVQA to classify if a slide is containing diagrams, figures, and images. Classification successful rate is at 0.8. Further implementation is needed to create bounding boxes to classify multiple elements and create multiple labels on one slide.

```
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Classes: ['Diagram', 'Figure', 'Image']
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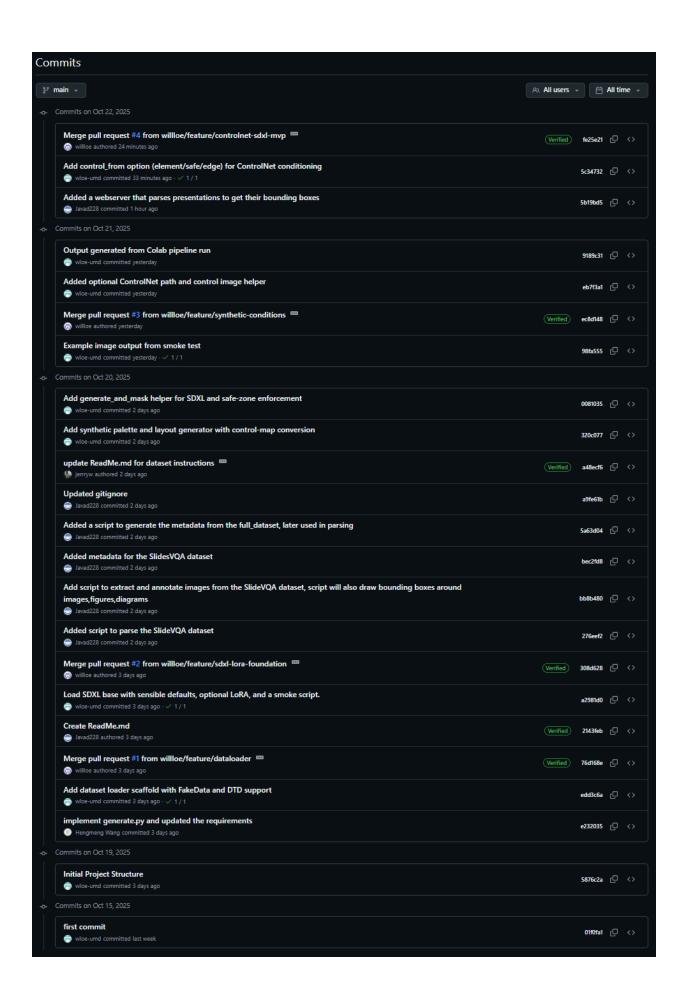
Javad: Generated metadata for the SlideVQA dataset using a script, which will later be used for parsing. Created a script to parse the dataset and extract and annotate images from the dataset with bounding boxes which are shown on each slide. Started implementing a web server that parses presentations to compile training data.

Diffusion Model:

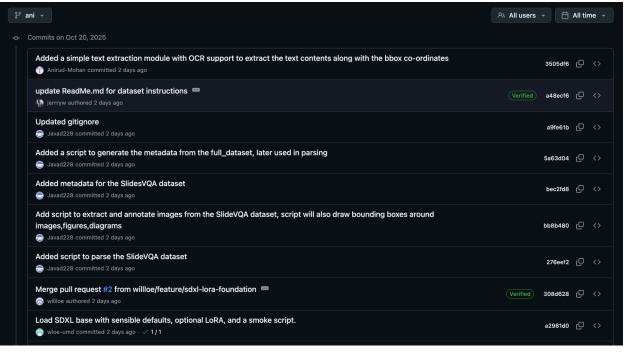
Hengmeng: Implemented structured layout data (like positions and types of titles, images, logos) into a 4-channel control tensor that guides diffusion models during image generation. It encodes element positions, class types, layering (z-order), and reading order into spatial maps that condition the model to respect layout constraints, while also creating "safe zone" masks for areas where the model can generate freely without layout restrictions. Essentially, the "generate.py" is a preprocessing pipeline that translates human-readable design specifications into the continuous tensor format needed by ControlNet-style diffusion models to generate images that follow specific layout structures.

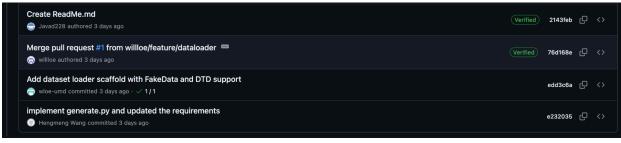
William: So far, I have set up a working SDXL-based generation stack with layout awareness and reproducible smoke tests. I defined and validated compact JSON Schemas for slide layouts and color palettes. I also built a synthetic data module that produces realistic, schema-valid layouts/palettes so I can test without external datasets. On the model side, I wired up sdxl.py to load Stable Diffusion XL with optional LoRA weights and sensible CPU/GPU defaults. I implemented infer.py to run generation and apply a safe-zone postprocess that keeps text regions neutral for readability. I also added an optional ControlNet path with a control_from switch (element/safe/edge) so I can experiment with which conditioning signal works best. The pipeline now saves the exact control image, raw generation, masked output, and an overlay for quick inspection.

	ipts (smoke_sdxl_load.py, smoke_synthetic.py, (CPU skips heavy gen) and on Colab (GPU) using the
GitHub Activity Snapshot	
Main branch:	



ani:





feature/graph-classification

