

AI for Automation of Modeling and Simulation (M&S) Workflows

Sponsored by General Dynamics Electric Boat

Senior Design CSE Team #46 & ECE Team #37

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Professor Jinbo Bi and Professor Park Sung-Yeul

Agenda



ENGINEERING
CHALLENGE



OBJECTIVE



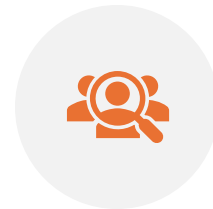
CSE & EE TEAM
RESPONSIBILITIES



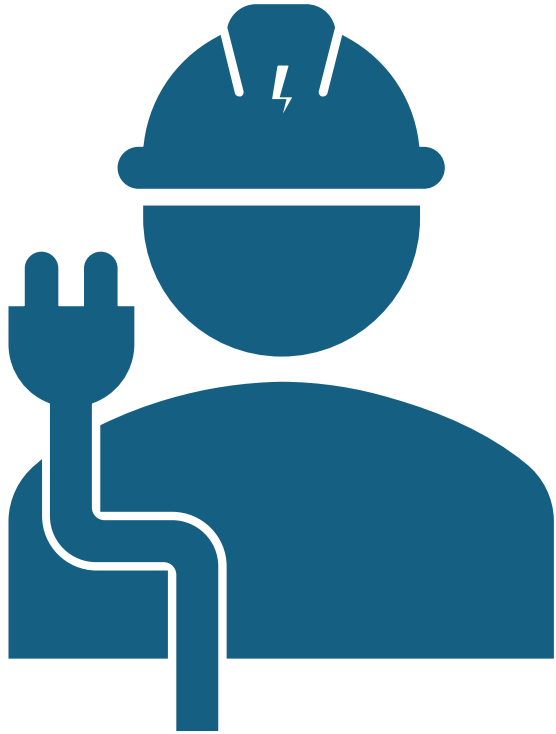
SYSTEM PIPELINE



WORKING DEMO

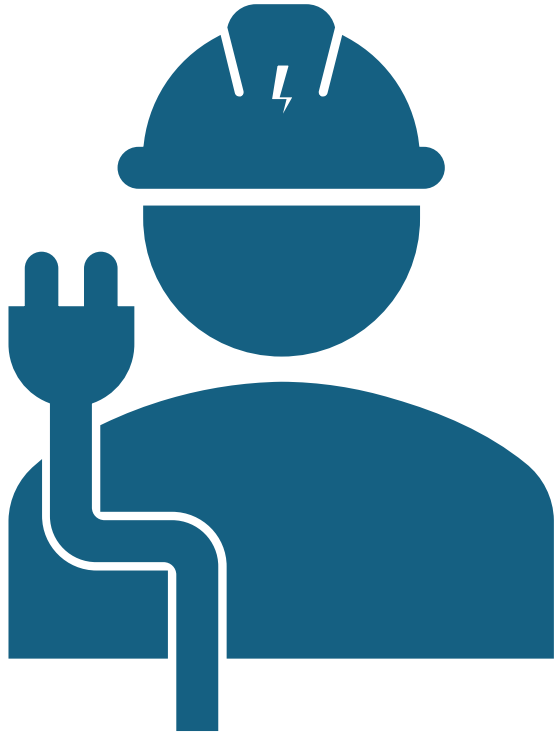


FUTURE
DEVELOPMENTS



Engineering Challenge

- Engineers at General Dynamics Electric Boat (GDEB) heavily rely on computer modeling and simulation (M&S) tools, like MATLAB/Simulink for design, testing, and troubleshooting.
- Tasks involving motor drives, inverters, DC-DC converters, and KPI analysis require running many repetitive simulations.
- Creating and modifying models is a manual process that requires engineers to write scripts, configure parameters, and interpret simulation results; tasks that consume valuable time and expertise.



Limitations

- GDEB is unable to provide the team with their data and models for training purposes
- This means we have to use open source Simulink models when training and testing
- Computational resources are limited as we need a lot of RAM to train and run models locally

Objectives



The project aims to leverage MATLAB/Simulink to improve the efficiency of M&S workflows.



The project will develop an AI-based tool to automate simulation tasks such as script generation, parameter configuration, and optimization.



The system will enable users to generate and run simulations using natural language prompts.



Automating repetitive tasks will improve engineering productivity.

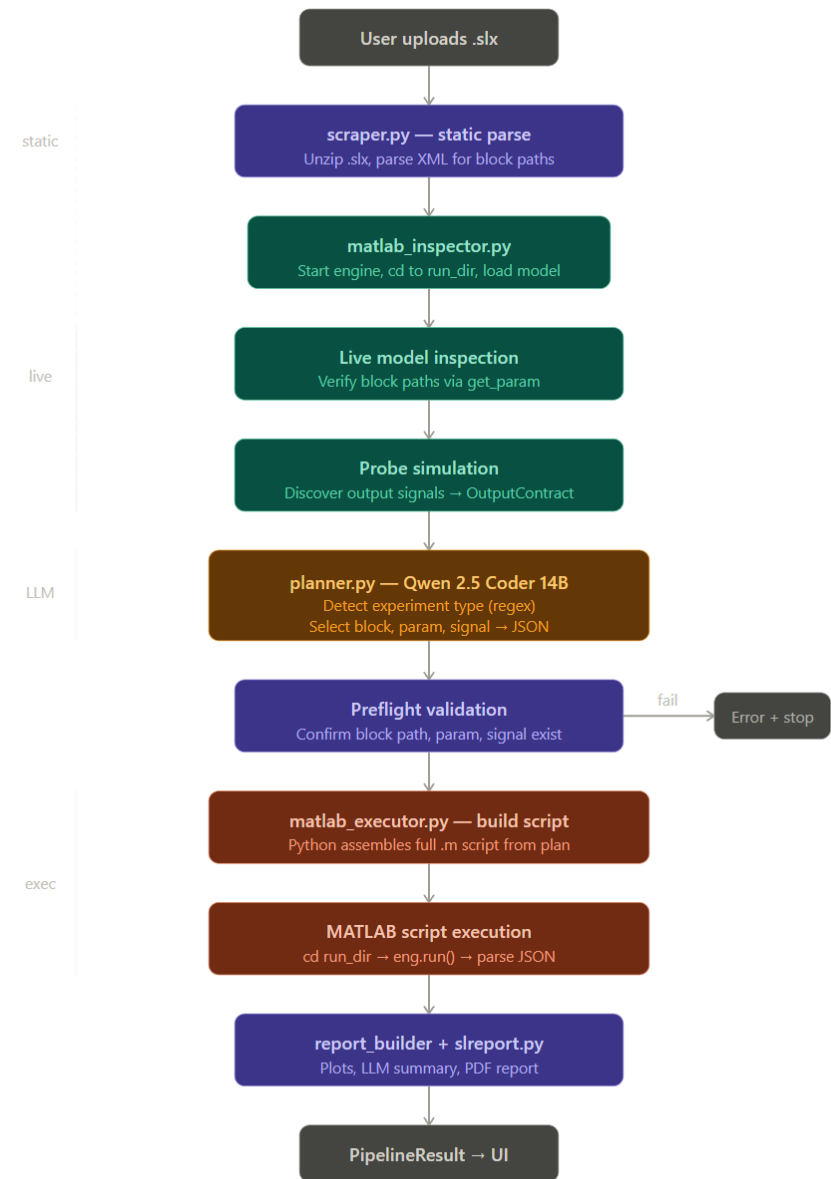
CSE Team Responsibilities

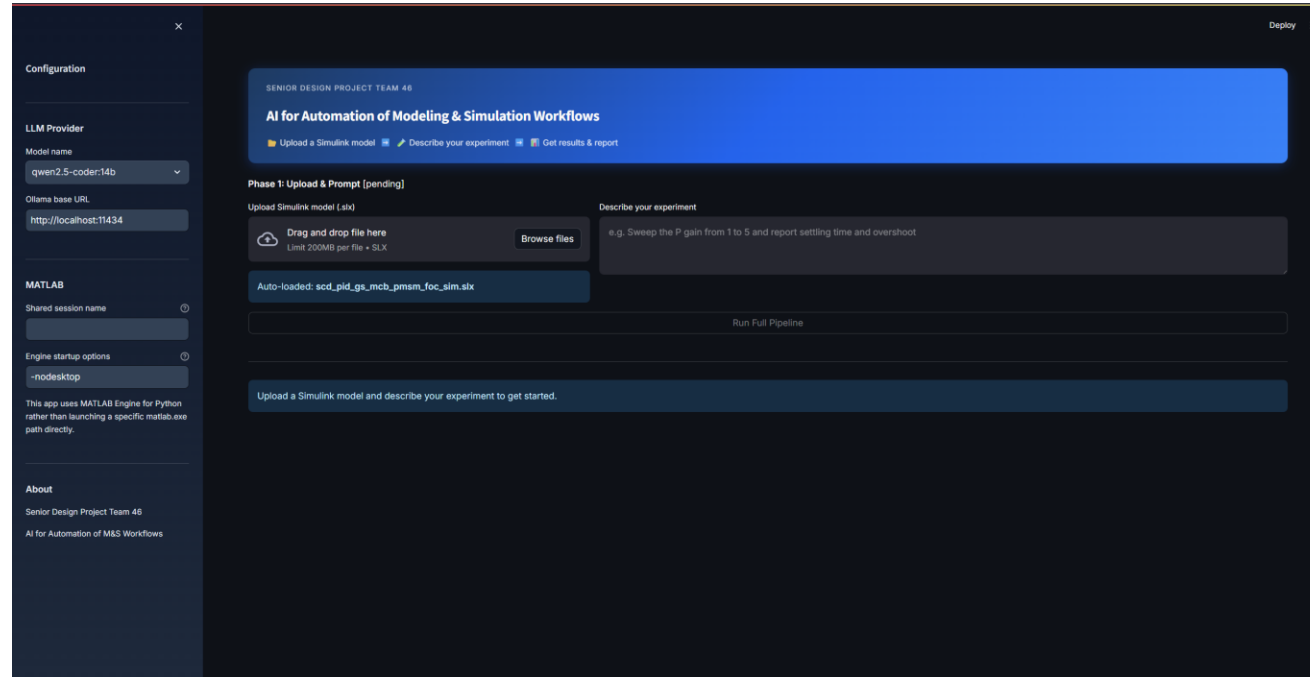
- Focused on developing the LLM and identifying the resources needed to support its implementation
- Model Selection, training, and fine-tuning
- Python GUI development
- Simulation Report & Analysis

ECE Team Responsibilities

- Assisted with training and developing an understanding of the various models involved
- Collected, cleaned, and provided the CSE team with data and parameters for use in LLM training
- Created prompts and scenarios for testing and training purposes
- Validated simulation report output
- Manual verification of LLM output and data analysis

System Pipeline



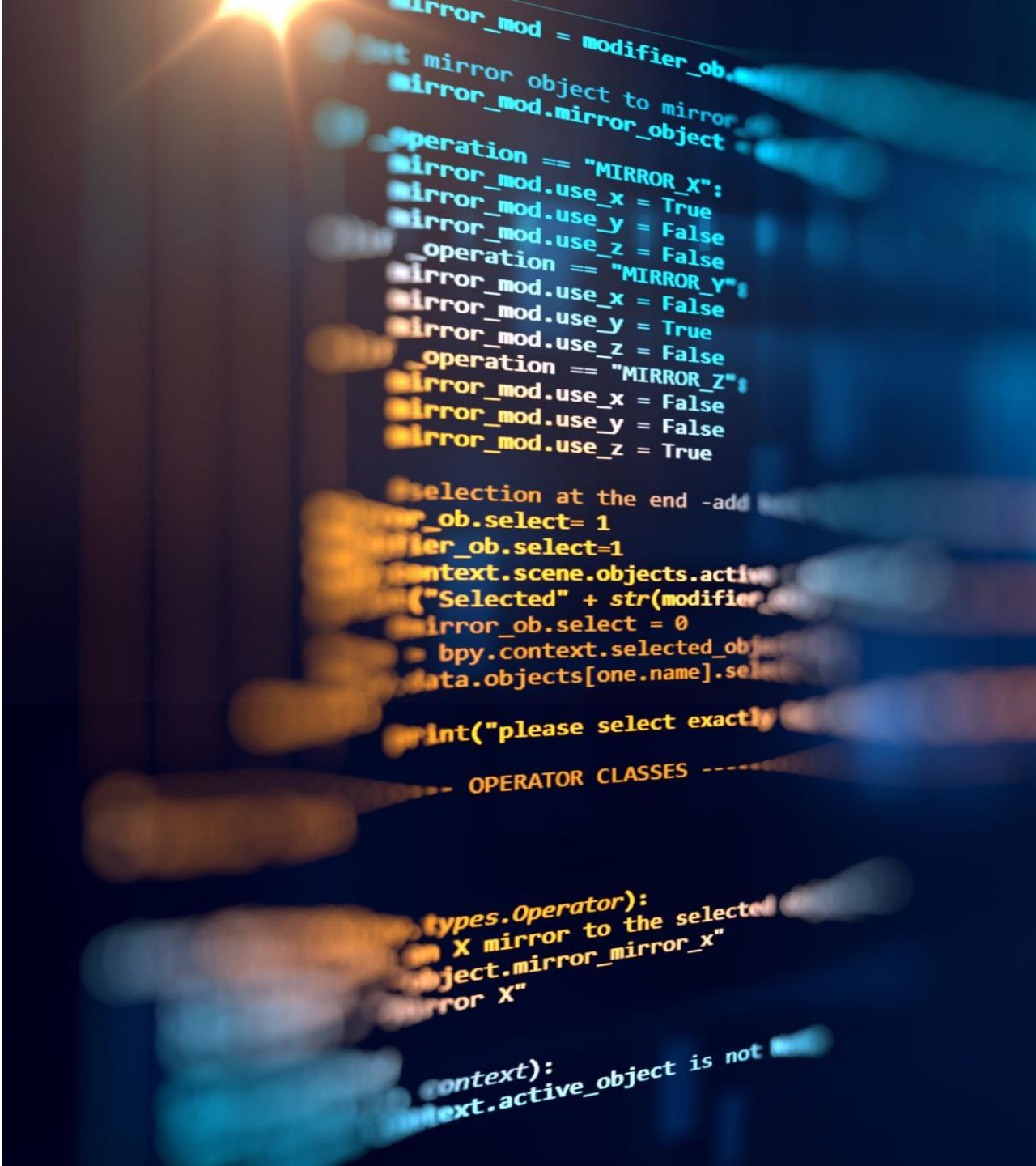


- The UI also prompts the user to upload a natural language prompt which is meant to be an experiment/analysis question about the Simulink model
- Upon launching our UI with Streamlit the user is prompted to upload a Simulink model in the form of a .slx file

User Interface

Scraper

- Opens .slx as ZIP → reads XML files
- Reconstructs full block hierarchy
- Extracts blocks, parameters, and connections
- Builds model summary for LLM context in natural language





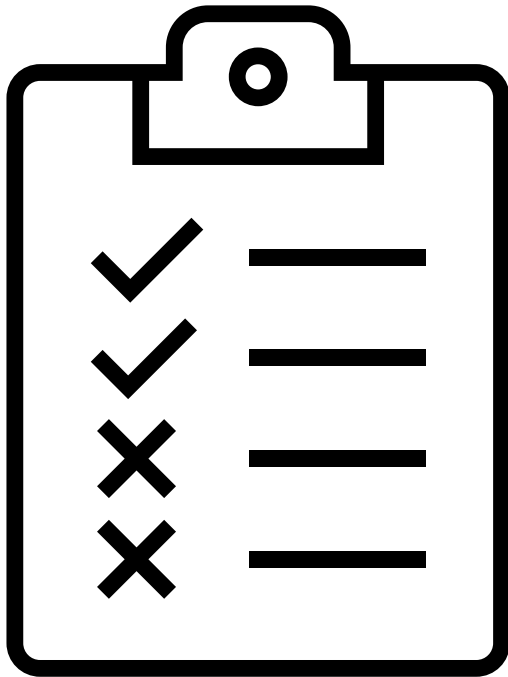
MATLAB Inspector

- Launches or connects to MATLAB engine
- Copies model to run directory
- Sets working directory
- Loads model into MATLAB

MATLAB Inspector – Verify Model

- Confirms block paths using MATLAB (`find_system`)
- Identifies tunable blocks (e.g., PID, Gain)
- Detects masked parameters
- Extracts current PID values





MATLAB Inspector – Probe Simulation

- Runs model once to detect outputs
- Checks outputs in priority order:
- logstdout → yout → simlog → workspace
- Builds OutputContract (signal names + indices)



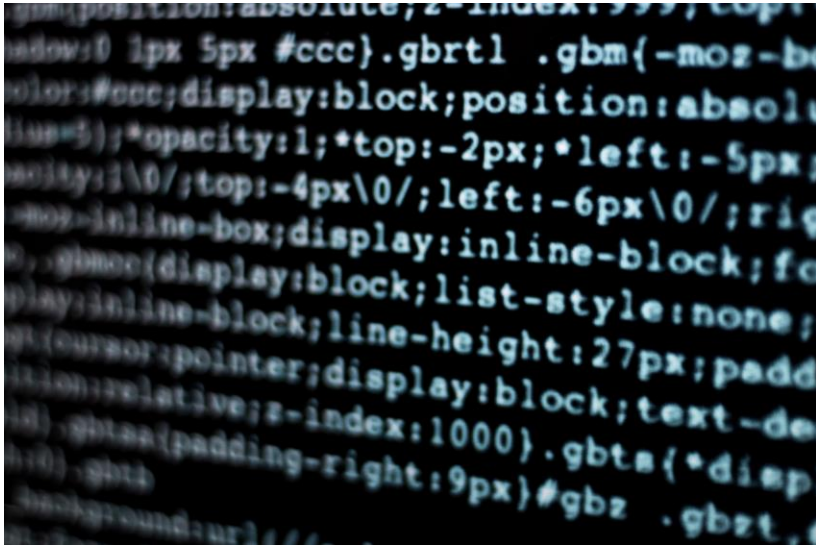
Planner

- Detects experiment type from prompt
- Filters valid tunable parameters
- Sends structured prompt to LLM
- LLM returns JSON experiment plan
- Python validates and finalizes plan



Planner - Preflight Validation

- Confirms block exists in model
- Verifies parameter is valid
- Ensures signal exists in outputs
- Stops early if errors found



MATLAB Executor

- Generates .m script entirely in Python
- Selects correct signal extraction method
- Adds loop (for sweeps) or single-run logic
- Appends metric calculations + helper functions
- Wraps script in try/catch

MATLAB Executor - Report Builder



- Saves and runs script in MATLAB
- Extracts `sdp_result` struct
- Converts results \rightarrow JSON \rightarrow Python
- Builds structured `RunResult`
- Optional metric refinement

Report Output

- Generates plots (time + metrics)
- LLM writes experiment summary
- Creates PDF report (MATLAB tools)
- Returns results to UI for display/download



Demo

The screenshot shows a web application interface with a dark theme. On the left is a sidebar with a 'Configuration' section containing:

- LLM Provider**
 - Model name: qwen2.5-coder:14b
 - Ollama base URL: http://localhost:11434
- MATLAB**
 - Shared session name: [empty]
 - Engine startup options: -nodesktop
- About**
 - Senior Design Project Team 46
 - AI for Automation of M&S Workflows

The main content area features a blue header with the title 'AI for Automation of Modeling & Simulation Workflows' and navigation links: 'Upload a Simulink model', 'Describe your experiment', and 'Get results & report'. Below this is a 'Phase 1: Upload & Prompt [pending]' section with two columns:

- Upload Simulink model (.slx)**: Includes a 'Drag and drop file here' area (200MB limit) and a 'Browse files' button. An 'Auto-loaded' file 'scd_pid_gs_mcb_pmsm_foc_sim.slx' is shown below.
- Describe your experiment**: A text area containing the prompt: 'Run the baseline speed-tracking case and explain what happened in the reference-versus-feedback graphs like an electrical engineer. Focus on tracking quality, overshoot, settling, ripple, and any control concerns.'

A red progress bar at the bottom of the main area is labeled 'Run Full Pipeline'. At the very bottom, there is a dark blue bar with the text 'Upload a Simulink model and describe your experiment to get started.'

Output Verification

- While the reports do show promise, the EE team needed to verify manually the accuracy of the reports generated by the LLM
- The reports are generated with the following categories in order:
 - Experiment Overview, Local LLM Analysis, Model Architecture, Simulation Results, Plots, Generated MATLAB Script
- In order to verify these results, a script was written to check and verify the results of the baseline run.
 - If the baseline is able to run properly, the other experiments can run and compare to the baseline properly

Call with Laing, Halmar

2026-04-06 21:33 UTC

Recorded by

Berry, Elizabeth

Future Developments

01


Build a bigger dataset of specific Electric-Boat models and train off those.

02

Use a bigger parameter model.

03

Add more agents to split up work to improve accuracy.



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