

TANGRAMFLEX[®]



LLM Co-pilots for Domain Specific Modeling Languages

AI4SE & SE4AI Research and Application Workshop
September 18, 2024

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Some Baseline Terminology

- ▶ **Artificial Intelligence (AI)** - *computer systems performing tasks that normally require human intelligence*
- ▶ **Generative AI** - *AI capable of generating text, images, or other data using generative models, often in response to prompts*
- ▶ **Large Language Models (LLMs)** - *AI that is capable of general-purpose language generation*
- ▶ **GPT** - *A general purpose LLM created by OpenAI*
- ▶ **ChatGPT** - *A user-friendly chat prompt interface that uses GPT created by OpenAI*



System Models and Generative AI in DoD Context

- ▶ All of the services are pushing to design, develop, test, and deploy faster
- ▶ Generative AI, which burst into mainstream consciousness in late 2022, has the potential to make lots of things go faster
- ▶ All of the services are pushing for increased digital engineering, digital transformation, and government-owned architectures and interfaces

As a DoD industry, we need a way to “go faster” and lean into digital engineering, and generative AI is a key enabling technology to make this happen



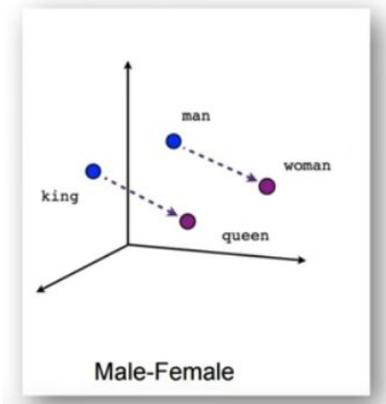
Generative AI for Making System Models

- ▶ Currently available generative AI technology is great at producing text for paper writing, emails, proposal writing, etc.
- ▶ Generative AI also holds great potential for accelerating digital systems modeling
 - ▷ When used effectively, generative AI can significantly reduce the human labor required to create digital models
- ▶ To fully achieve the promise, generative AI must first overcome significant scaling challenges



Generative AI for System Models - Scaling Challenges

- ▶ Current generative AI technology suffers from a correctness/accuracy problem
 - ▷ LLMs can't tell if what they've produce is correct or not
 - ▷ Incorrect outputs are commonly called “Hallucinations”
 - ▷ This is due to the way LLMs work

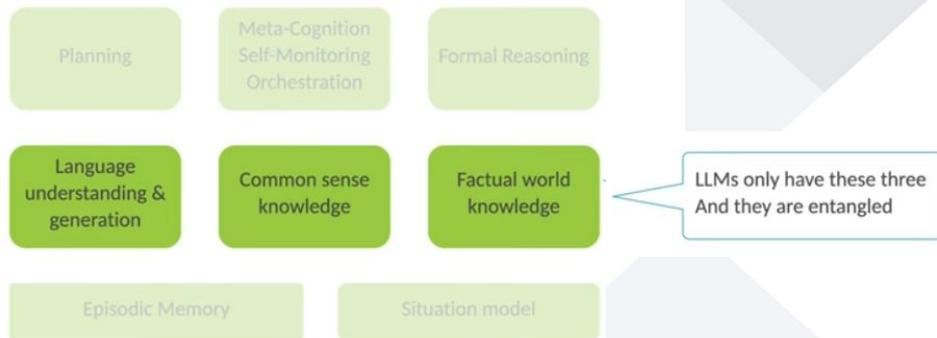


Dr. John Launchbury
Tech Talk - “The
Trajectory of AI”

[https://www.youtube.com/
watch?v=OPFSXSZmeOQ](https://www.youtube.com/watch?v=OPFSXSZmeOQ)



The geometry of words



The elements of human intelligence

Generative AI for System Models - Scaling Challenges

- ▶ The hallucination problem is compounded when generating general purpose modeling specification
 - ▷ A wide-open specification gives the LLM many more options to generate results that may or may not be correct
 - ▷ Checking the results is usually a manual and tedious process
- ▶ Generative AI is most effective at producing text
 - ▷ Not all commonly used modeling tools adhere to a semantically-rigid textual standard, making it more difficult for the AI to produce accurate models



Domain Specific Modeling Languages (DSMLs) & Generative AI



What is a Domain Specific Modeling Language (DSML)?

- ▶ DSML is a language designed to model within a constrained domain
 - ▷ Intentionally is not general purpose
 - ▷ Can't model everything
- ▶ DSMLs typically have tighter syntax than general purpose languages
 - ▷ Well suited for precise specification enabling mathematical analysis, automatic code generation, configuration of systems, etc.
- ▶ DSMLs are increasingly used in industry and the DoD
 - ▷ Examples: VHDL, Terraform, YAML for CI/CD pipelines, MATLAB, AADL, Cryptol for cryptography, Flex for integration modeling



DSMLs Help Address LLM Scaling Challenges

- ▶ DSMLs are text based, making them ideal targets for generative AI
- ▶ LLMs are straightforward to train to produce syntactically accurate DSML code
 - ▷ DSMLs require less simultaneous context for the LLM than a broader general purpose language
 - ▷ Retrieval Augmentation Generation (RAG) can effectively augment general purpose LLMs without special model training
- ▶ LLMs can be augmented with tools built to work on DSMLs
 - ▷ Example: LLM could directly interact with a syntax checker and iterate until it has produced fully valid code



Examples From Our Work



Some Examples From Our Work

- ▶ Using generative AI to produce Flex, an integration modeling language, to create digital representations of message standards for code generation
- ▶ Using generative AI to produce a Tangram Pro component model that was automatically converted to a Cameo SysML model

Note: All experiments described here were done using AskSage with CUI/ITAR safe GPT 3.5 and 4 to prove their viability in a sensitive DoD environment



Example - Generating Message Library Code

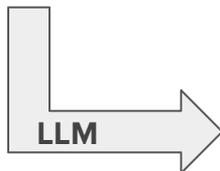
- ▶ Trained a RAG dataset on the Flex DSML by providing it the public website documentation on the language (<https://docs.tangramflex.io/flex/start>)
- ▶ Using a variety of message formats, provided GPT portions of the published standard (PDF), and asked it to generate a Flex DSML representation of the messages defined
- ▶ Provided the generated Flex DSML code to a C/C++ code library generator



Example - Generating Message Library Code

MESSAGE NUMBER/NAME: MT-031/ELEMENT HEARTBEAT						
SIGNAL NAME/SYMBOL	LOCATION			RANGE OF VALUE	UNITS	REMARKS
	WORD	# OF BITS	SBP			
Message Time Tag (MTT)	0-1	64	31		Seconds	Double precision word. Time is in seconds since midnight 1/1/1970 (Unix EPOCH).
Message Size (MS)	2	16	31	Fixed	Decimal	Set to 4.
Message Type (MT)	2	16	15	Fixed	Octal	Set to 031.
Reserved	3	8	31			
Reserved	3	8	23			
Enclosure ID (ENCLID)	3	8	15	0 to 255	Decimal	Set to 0. 0 = Not Used
Element ID (ELEMID)	3	8	7	Fixed	Decimal	Set to 37 (Radar A).

Documentation Table



```

// Element Heartbeat message struct
message struct ElementHeartbeat {
    // Time in seconds since midnight 1/1/1970 (Unix EPOCH)
    mtt : float64; // Message Time Tag (64 bits)

    ms : uint16; // Message Size (16 bits, fixed to 4)

    mt : uint16; // Message Type (16 bits, fixed to 031 octal)

    // Reserved fields (16 bits total, 8 bits each)
    reserved1 : uint8;
    reserved2 : uint8;

    encId : uint8; // Enclosure ID (8 bits, fixed to 0)

    elemId : uint8; // Element ID (8 bits, fixed to 37 for Radar A)
}

```

DSML Code



C++ Message Library



Example - Generating Message Library Code

- ▶ We measured *significant* time savings with this approach
- ▶ We gave the same challenge to 2 interns
 - ▷ Scenario: Develop a message library for STANAG-4586 (NATO standard for UAV payload and platform control) using Flex code generator
 - ▷ Neither had experience with STANAG-4586 or Flex or developing a message library
 - ▷ Both had similar academic and engineering experience
 - ▷ One intern did it without generative AI, other with generative AI
- ▶ Result: the intern with generative AI accomplished the task weeks faster

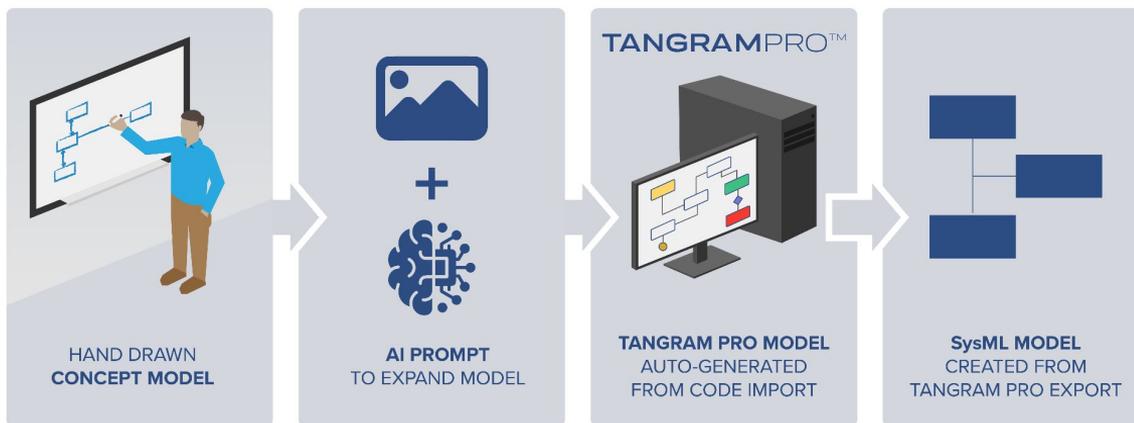


Example - Generating SysML Model from Sketch

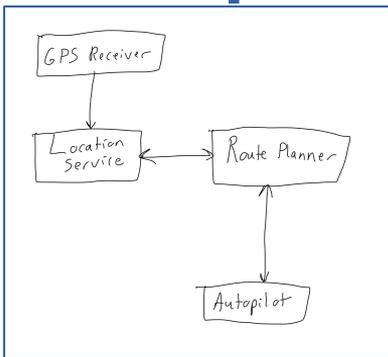
- ▶ Trained a RAG dataset on the component modeling specification language used under the hood by Tangram Pro Designer
- ▶ Provided a hand-drawn block diagram, and prompted the LLM to generate the corresponding component model
- ▶ Synchronized the component model into a SysML internal block diagram
- ▶ Repeated the test with prompts to the LLM to make modifications to the sketch



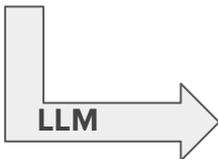
Example - Generating SysML Model from Sketch



Example - Generating SysML Model from Sketch



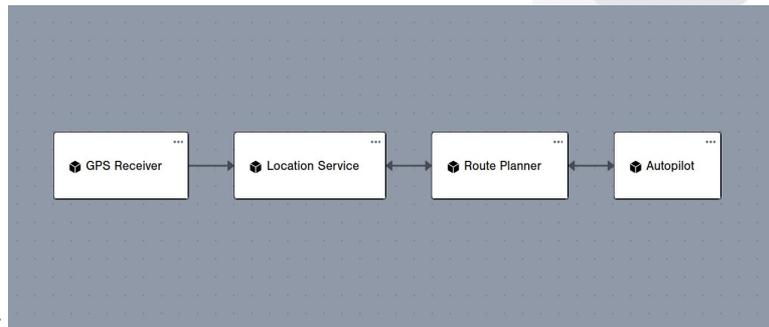
Hand Drawn Diagram



```
between "Route Planner" and "Autopilot."
{
  "name": "Autopilot Integration System",
  "description": "A system that integrates GPS data with an autopilot through a route planner with bi-directional connections.",
  "branches": {
    {
      "name": "master",
      "description": "Main integration branch for the autopilot system with bi-directional connections.",
      "head": {
        "id": "head-id-123",
        "name": "head-name-autopilot-system",
        "type": "PROJECT",
        "description": "Head of the master branch for the autopilot integration system.",
        "source_repo_url": "",
        "source_repo_revision": "",
        "properties": null,
        "subcomponents": {
          {
            "id": "gps-receiver-id",
            "name": "GPS Receiver",
            "type": "BASIC",

```

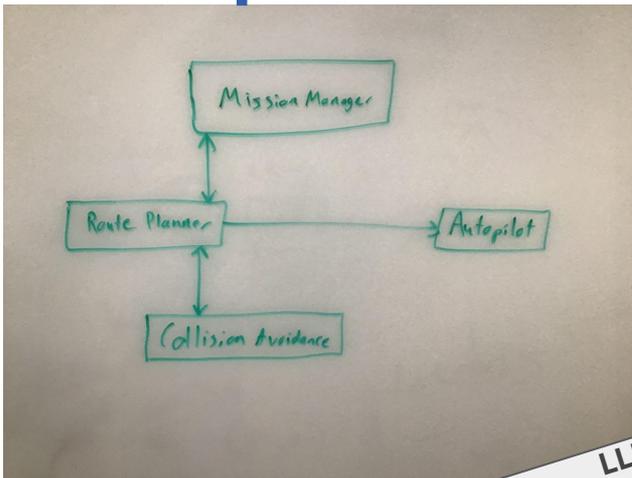
Generated DSML Code



Digital System Model



Example - Generating SysML Model from Sketch



Whiteboard Diagram



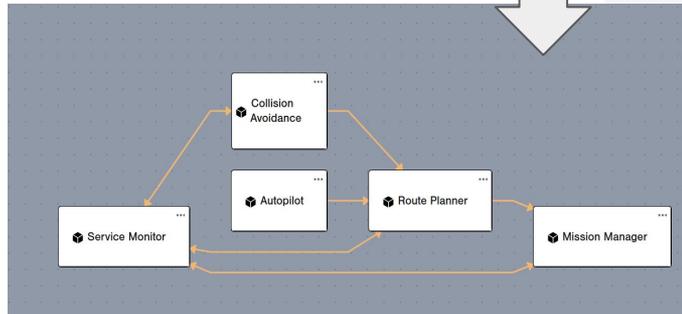
You: This picture describes a Tangram Pro project. Convert this to a Tangram Pro JSON file with the following changes. Add an additional component called Service Monitor that has a bidirectional connection to Mission Manager, Route Planner, and Collision Avoidance.

Based on the image provided, I will create a Tangram Pro JSON file for the project, including the additional "Service Monitor" component with bidirectional connections to "Mission Manager", "Route Planner", and "Collision Avoidance".

```
{  
  "name": "Enhanced Mission Control Project",  
  "description": "A project with a Service Monitor that has bidirectional connections to Mission Manager, Route Planner, and Collision Avoidance.",  
  "branches": [  
    {  
      "name": "Service Monitor",  
      "description": "A component that monitors the system and provides feedback to the Mission Manager, Route Planner, and Collision Avoidance.",  
      "type": "Component",  
      "properties": {}  
    }  
  ]  
}
```

Prompted changes and generated DSML code

DSML Import



Digital System Model



Example - Generating SysML Model from Sketch

- ▶ The LLM plus DSML tools were able to turn a hand drawing/whiteboard sketch into a digital model block diagram in minutes
- ▶ No straightforward path that we knew of to generate a SysML model from LLM without help of DSML/external tooling
- ▶ These results would likely hold true with voice/audio recordings as well
 - ▷ It should be possible to explain a design and have an LLM generate a digital model of the design, via an appropriate DSML



Final Thoughts for Today



Final Thoughts for Today

- ▶ DSMLs can address the gap between current generative AI technology and current systems engineering modeling practices
- ▶ When used effectively, LLMs that produce DSML output for the purposes of modeling can provide significant time savings over manual model creation
- ▶ Models still must be verified for accuracy, but verifying DSML models first can produce more accurate models in the end product format
- ▶ Let's lean into the technology we have today to go faster!



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