



Tracil: AI-Powered Traceability Tool Across CDISC Standards

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Meet the Speakers

Kexin Guan

Title: Scientist, Statistical Programming

Organization: Merck & Co., Inc.

Kexin Guan is a Statistical Programmer at Merck & Co. Inc., where she has been part of the Oncology Early Development group since December 2022. She holds a Master's degree in Biostatistics and Bachelor's degree in Applied Mathematics and Statistics.



Junze Zhang

Title: Scientist, Statistical Programming

Organization: Merck & Co., Inc

Junze Zhang is a Scientist at Merck & Co. Inc., supporting early oncology statistical programming. He earned his Master's in Computer Engineering from NYU and Bachelor's in Computer Science from Oregon State University.





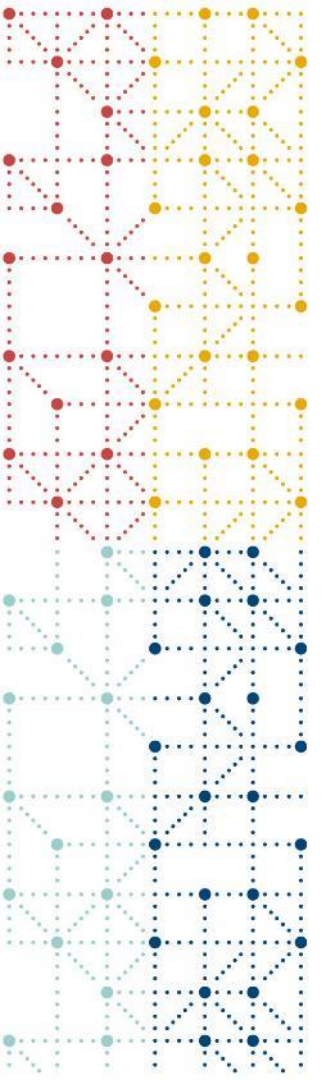
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- *The views and opinions expressed in this presentation are those of the author(s) and do not necessarily reflect the official policy or position of CDISC.*
- *The views and opinions expressed in this presentation are those of the author(s) and do not necessarily reflect the official policy or position of Merck & Co., Inc.*



Agenda

1. Background & Motivation
2. Demo
3. App Structure
4. Backend AI Workflow
5. Summary, Lessons Learned & Future Steps



Background & Motivation



The Traceability Problem

- Clinical data flow is **complex**, spanning **multiple** systems and **silos**.
- Each layer (Protocol → CRF → SDTM → ADaM → TLFs) adds transformation logic.
- Manual tracing lineage = **heavy review time**.

Our Project Goals

- **Automate Lineage Inference**
 - Extract relationships from existing metadata without manual mapping.
- **Support CDISC Standards End-to-End**
 - Protocol (USDM) ↔ CRF ↔ SDTM (Define.xml) ↔ ADaM (Define.xml) ↔ TLF (ARS)
- **Provide Explainable AI Results**
 - Every link comes with a natural-language justification.
- **Deliver Human-Friendly Visualization All in One Place**
 - Interactive graphs for regulators, programmers, and statisticians.
- **Explore Data with Natural Language**

User Flow



Inputs

- Protocol (USDM / PDF)
- CRF (annotated CRF)
- SDTM / ADaM
(Define.xml / spec.xlsx)
- TLF (RTF or ARS/ARD)



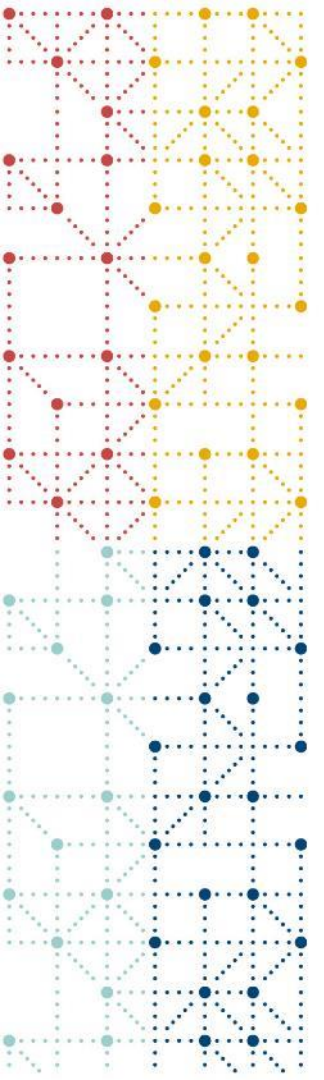
Processing

- Preprocessing
- AI model analysis
- Lineage image
generation



Outputs

- Interactive Lineage
Graph



Demo

Upload & General UI

DEMOS

ADaM

The screenshot shows a web browser window with the URL `localhost:3000`. The application is titled "Tracil - Clinical Data Lineage". On the left sidebar, under the "ADaM" section, the "ADSL" dataset is selected. Below it, a list of other datasets is visible: ADAE, ADLBC, ADLBH, ADLBHY, ADQSADAS, ADQSCIBC, ADQSNPIX, ADTTE, and ADVS. Under the "SDTM" section, a list of datasets is visible: TA, TE, TI, TS, TV, DM, and SE. The main content area displays the "ADSL" dataset, which has 25 variables. The source files are listed as "define.xml (primary)". A grid of 20 variable cards is shown, arranged in 4 rows and 5 columns. The variables are: ARM, TRT01P, TRT01PN, TRT01A, TRT01AN, AGEGR1, AGEGR1N, AGEU, RACE, RACEN, SEX, ETHNIC, SAFFL, ITTFL, EFFFL, COMP8FL, COMP16FL, COMP24FL, DISCONFL, DSRAEFL, BMIBLGR1, DURDSGR1, DCDECOD, and DCREASCD.

Search variables...

ADSL ADSL

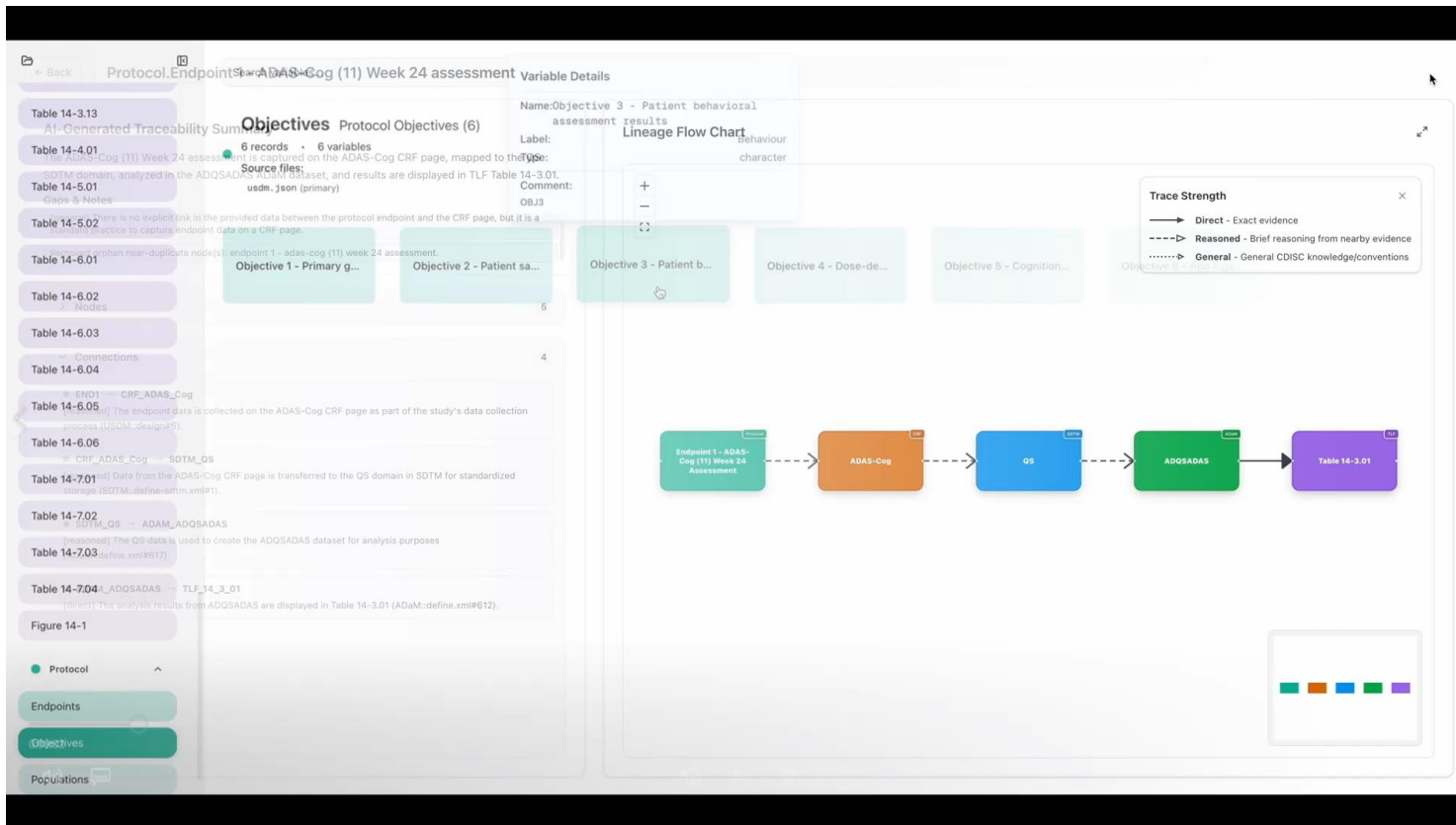
- 25 variables

Source files:

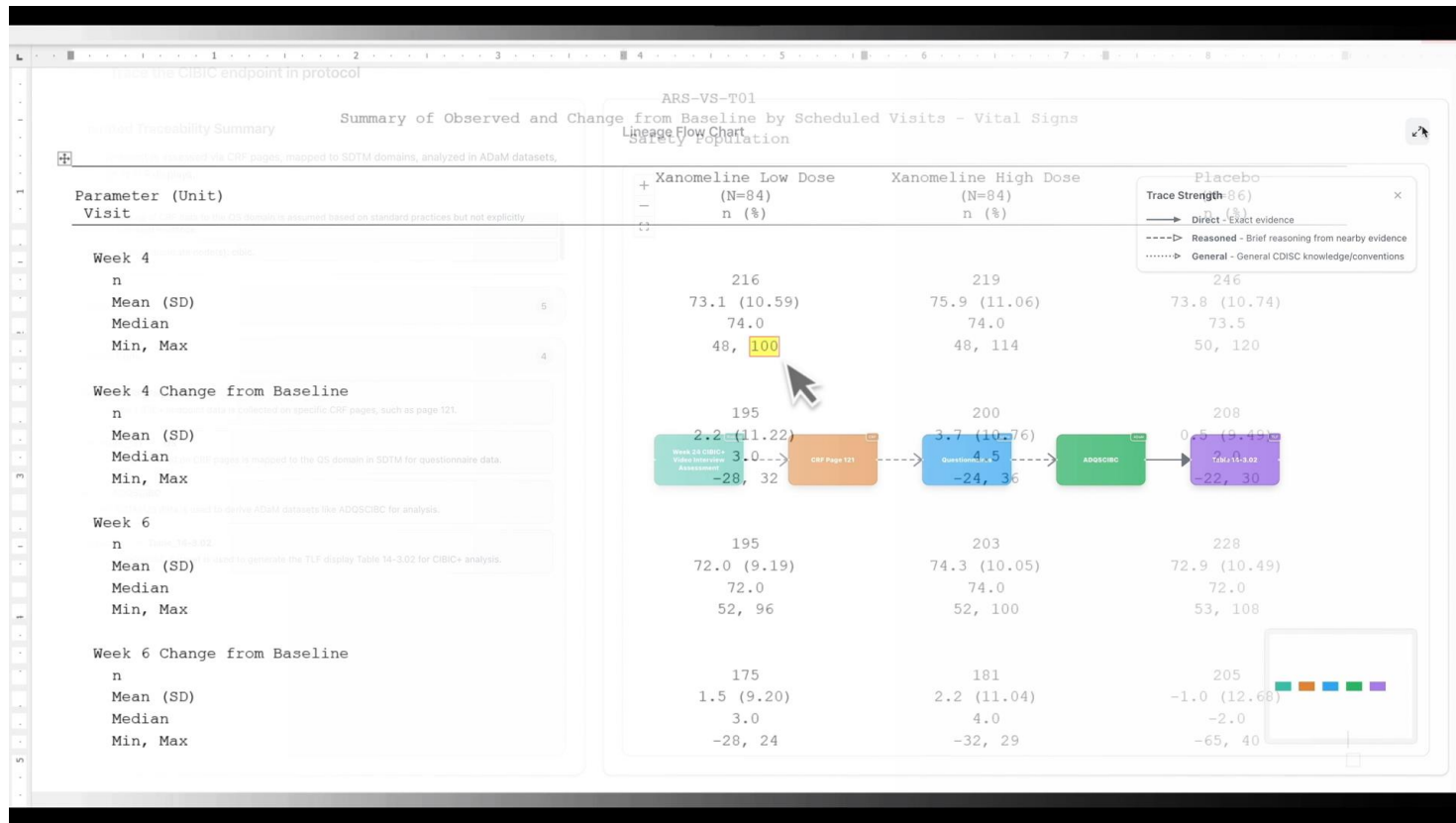
- define.xml (primary)

ARM	TRT01P	TRT01PN	TRT01A	TRT01AN	AGEGR1	AGEGR1N
AGEU	RACE	RACEN	SEX	ETHNIC	SAFFL	ITTFL
EFFFL	COMP8FL	COMP16FL	COMP24FL	DISCONFL	DSRAEFL	DTHFL
BMIBLGR1	DURDSGR1	DCDECOD	DCREASCD			

Objectives



TLF Details with ARS

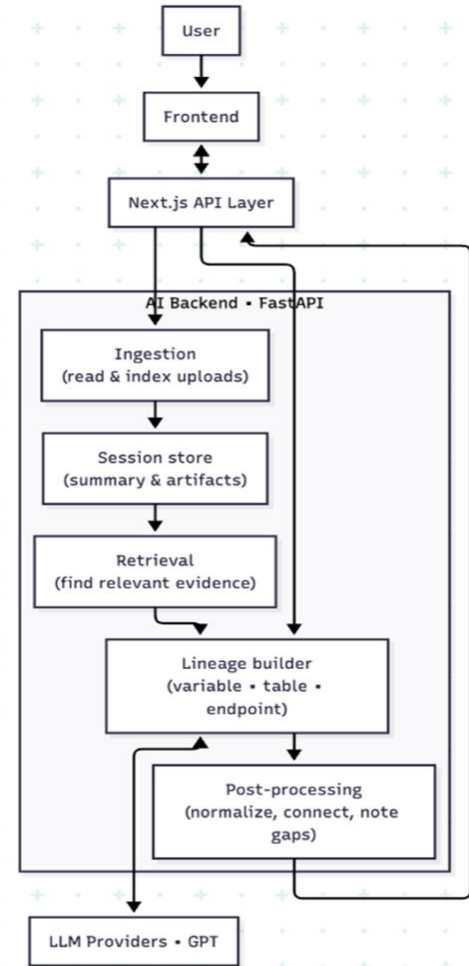




App Structure

App Structure

- Frontend: Javascript with React/Next.js
- Backend: FastAPI
- AI core: Python AI Engine
 - GPT-4o (mini)
 - Text-embedding-3-model from OpenAI





Backend AI Workflow



How Tracil's Backend Works

- **Pre-processing:** Python code to convert structured (XML, JSON) and semi-structured (PDF/CRF) inputs into a unified, machine-readable JSON file
- **LLM Reasoning:** passes the inputs to an LLM reasoning layer to infer variable derivations, dependencies, and gaps, returns a JSON graph
- **Post-processing:** standardize the JSON lineage graph so that it can be visualized interactively in the frontend UI



Pre-Processing: Parsing & Normalization Layer

Input formats supported: Specification (.xlsx), define.xml, aCRF, ARS JSON, Protocol PDF, USDM, and TLFs (RTF)

FastAPI endpoint /process-files:

- Extracts variable metadata, derivation notes, and dataset context
- Normalizes the information extracted into unified schema of names, domain tags, and relationships information

Why important:

- Provides the frontend with structured data for visualization and user interaction.
- gives LLM a clean, consistent input so it can reason semantically



LLM Lineage Builders

Three Routes by Target

- Detects what you're tracing: protocol endpoint, ADaM/SDTM variable, table/cell; sends it to the right builder

Gather the Evidence

- Collects all supporting pre-processed metadata from the session (aCRF index, protocol text, unified JSON returned by /process-files API, etc.)
- These documents form the “evidence base” for reasoning

Find What Matters (Chunk + Retrieve)

- Splits large documents into small readable sections
- Converts each into numerical “embeddings” so the AI can compare meanings
- Selects only the Top K (≈ 12) most relevant pieces to focus the analysis

Ask the AI Model

- Sends the selected context to GPT-4o for reasoning (mini model as fallback).
- The model returns a structured JSON lineage, showing each variable, link, and explanation.

Parse + Check

- Cleans the graph and validates to ensure it's proper JSON

Prompt Design

- **Clear Role Definition:**

- *System* = “Senior CDISC standards expert”; explicit backtrace & forward instructions; closed node types and canonical IDs (e.g., ADSL.AGE, DM.BRTHDTC)

- **Structured Thinking:**

- Each task follows a fixed schema (variable, endpoint, or table) so the AI always knows what format to produce

- **Evidence packing:**

- The AI reviews only the top relevant document sections, tagged as evidence, before reasoning; pins the target node (e.g., Target variable: ADSL.TRT01AN)

- **Explainable Results:**

- Every link in the lineage includes a short explanation starting with [direct], [reasoned], or [general], and cites where the information came from (define.xml, ARS, CRF, protocol)

- **Direction rules:**

- hard requirement to emit edges upstream → downstream only; forbid illegal shortcuts (e.g., ADaM → CRF)

- **Resilience:**

- chat called with temperature=0.0 + response_format={"type":"json_object"}; Built-in checks handle small formatting issues automatically, so results remain valid and reproducible



Post-Processing: Normalize, Validate, Connect

Normalize the Structure

- Add tags and standardize names so the graph speaks one language. (e.g., ADVS.AVAL → ADaM variable, VS.VSORRES → SDTM variable)
- Fix missing labels and edge directions for clear flow (source → result)

Validate the Connections

- Check each link has real evidence; flag gaps if missing
- Remove duplicates/orphans, and add short explanations

Connect Missing Pieces

- Re-query files when links are incomplete (e.g., ADaM variable with no SDTM parent)
- Ensure full trace: Protocol → CRF → SDTM → ADaM → TLF



Conclusion & Future Steps

Key Takeaways

- Tracil automates lineage across Protocol → CRF → SDTM → ADaM → TLF using AI reasoning
- Converts protocol endpoints, aCRF, define.xml, specifications, ARS/ARD into a **unified JSON schema**
- Provides **explainable AI outputs** with clear variable relationships
- Makes traceability **easy, fast, interactive, and follows CDISC standards.**



Limitations

- **Model Accuracy & Stability:** Same input can yield slightly different results due to the nature of LLM
- **Limited Data:** Few open, CDISC-compliant datasets restrict realistic fine-tuning
- **Validation Gap:** Need standardized validation methods to ensure reliable outputs

Future Steps & Vision

- **Model Expansion:** Test across different LLMs (GPT-5, Gemini, Claude, etc.)
- **Confidence Scoring + User Feedback:** Quantify AI certainty and learn from human corrections



Thank You!



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- Tracil GitHub Repository:
 - <https://github.com/1mgroot/Tracil>

