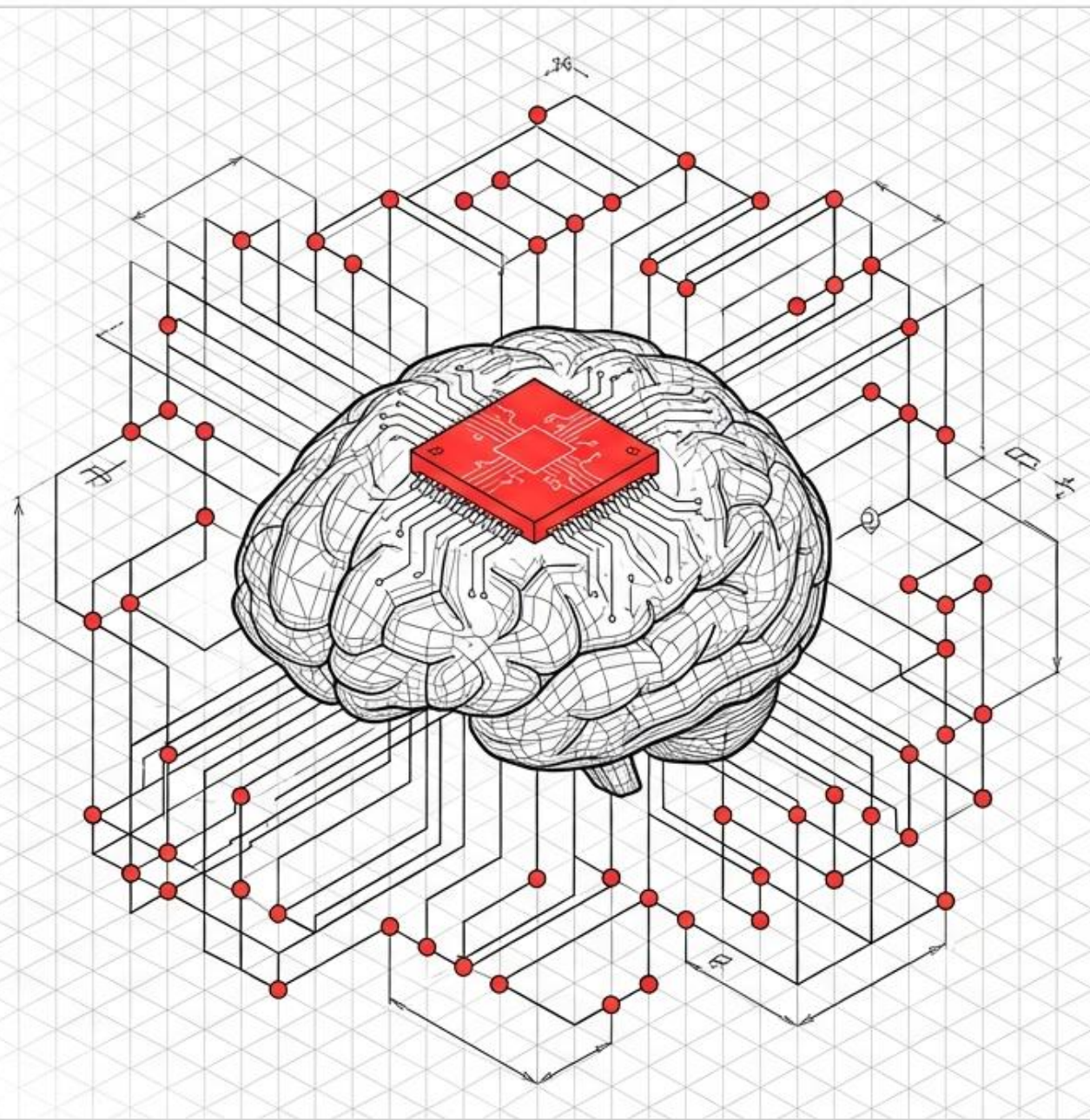


AI 101: BUILDING SUPERVISED MACHINE LEARNING MODELS

Operationalizing the 2026 Pivot to
AI-Driven Acceleration.

A hands-on technical bootcamp for
the Embers-to-Innovation Summit.

TRACK: TECHNICAL ENABLEMENT
SESSION: 01



THE 2026 MANDATE: SPEED AND PRECISION

THE HARD PIVOT

Moving from general “Energy Leadership” to demonstrable technical proficiency.



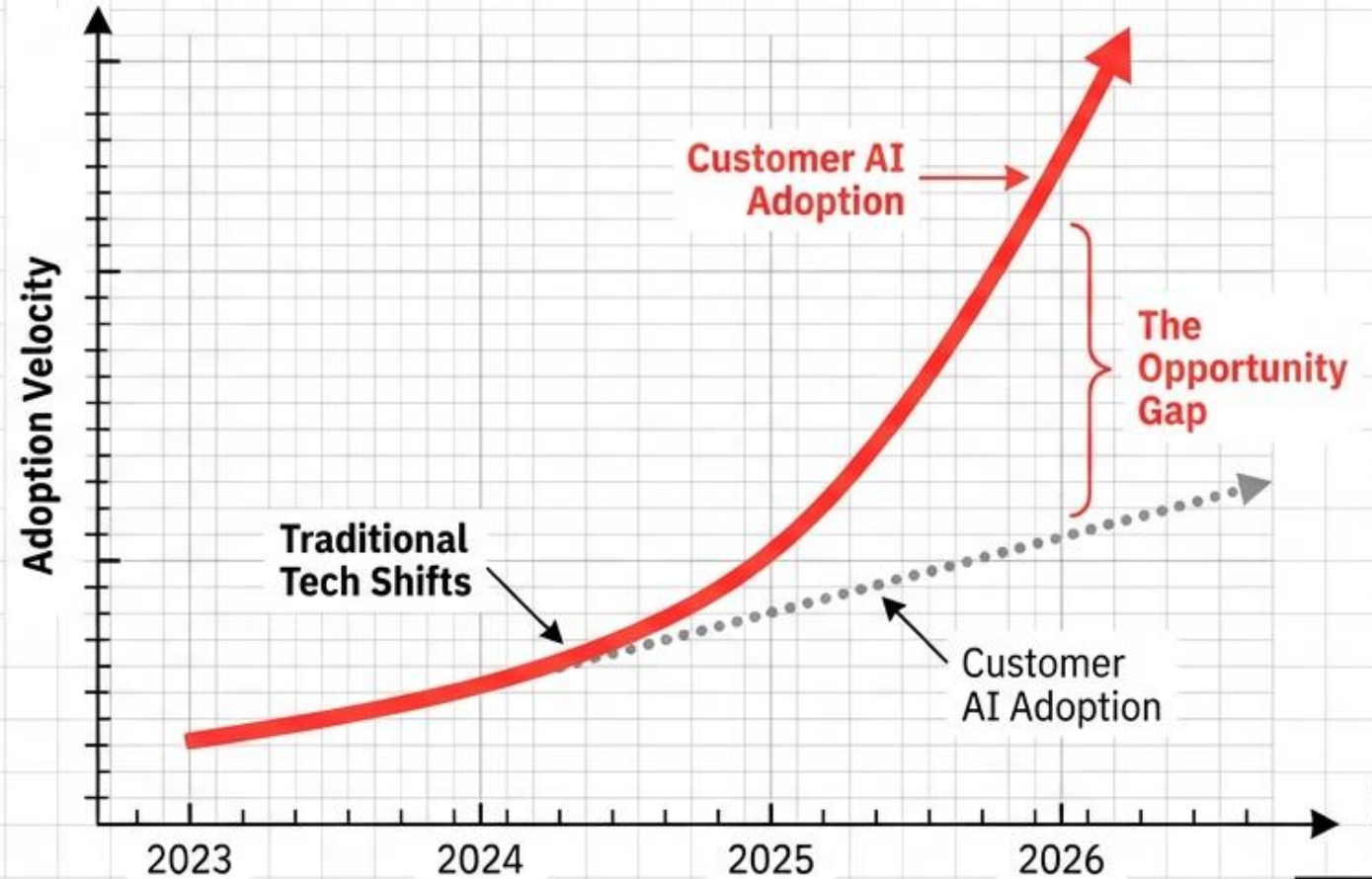
THE OBJECTIVE

Operationalize internal mandates to improve sales targeting and marketing speed.

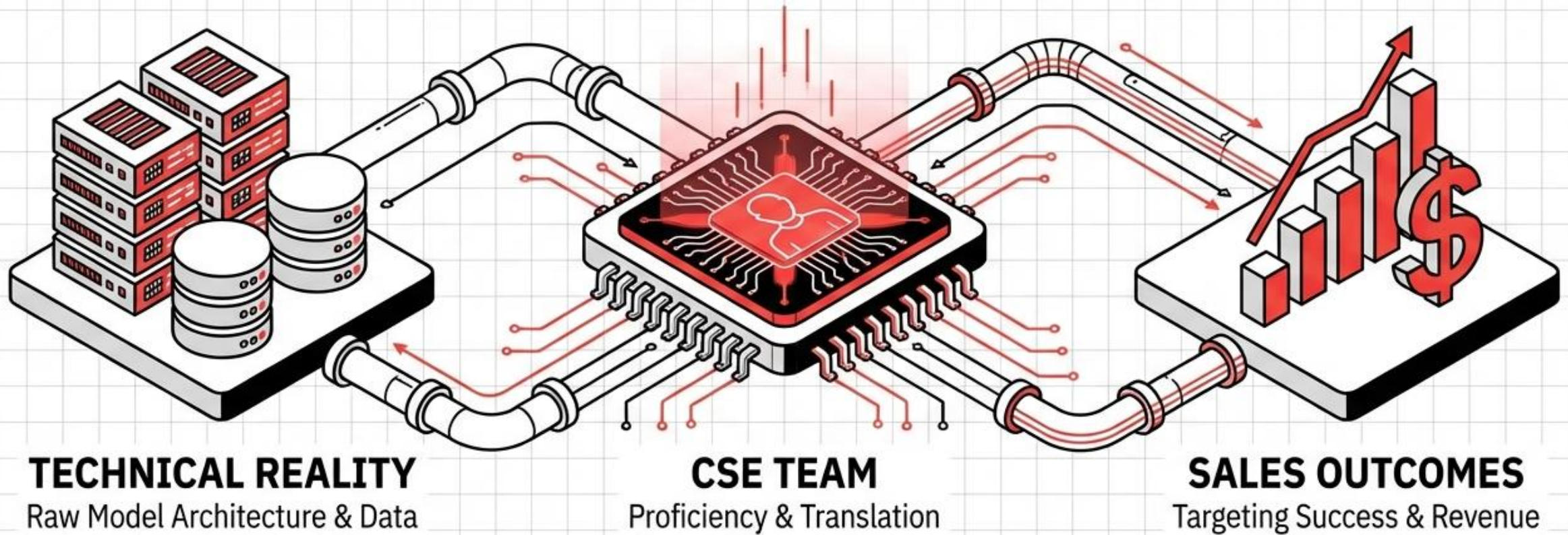


THE METRIC

Speed is the critical priority. We must enable the organization to move faster through AI adoption.



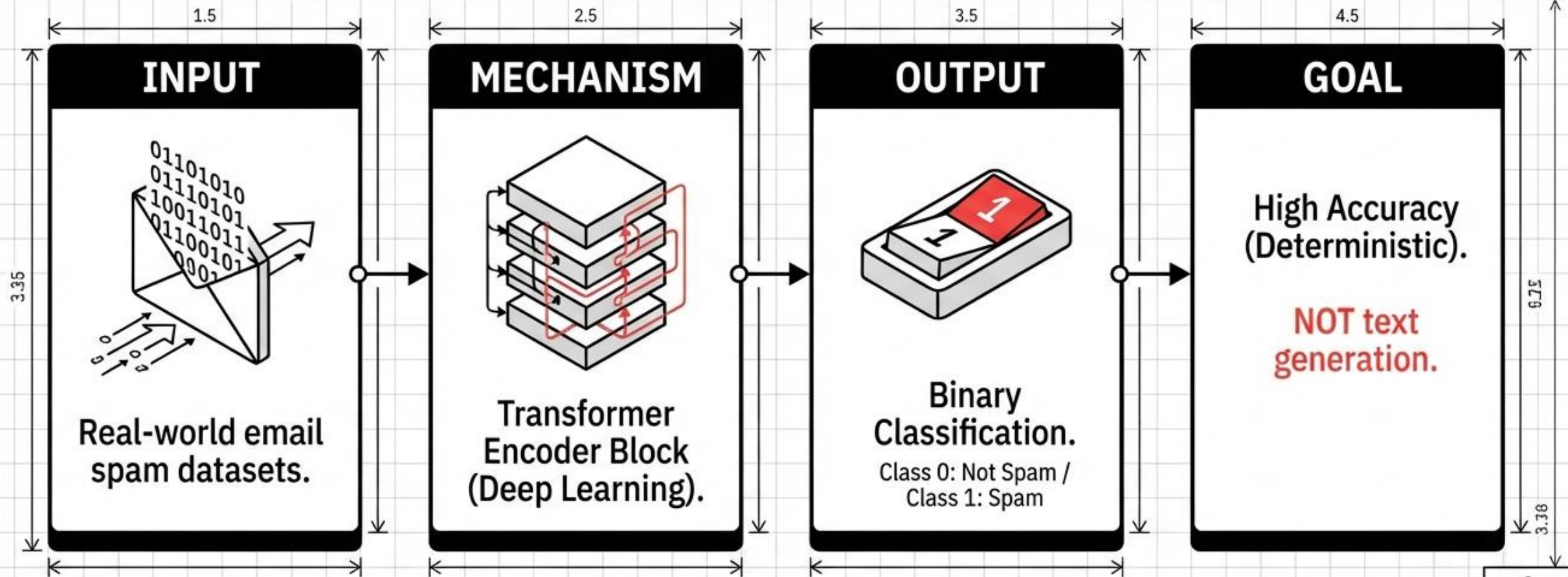
THE CSE TEAM ACTS AS THE NEURAL BRIDGE



Mission: You cannot sell or operationalize what you cannot build. Proficiency = Credibility.



LAB SCOPE: WHAT WE ARE BUILDING



SUPERVISED VS. GENERATIVE: THE GUARDRAILS

THIS LAB: SUPERVISED LEARNING



- **Deterministic** Predictions
- **Focus:** Classification Accuracy
- **Output:** Specific Categories (Spam/Not Spam)
- **Use Case:** Precise Sales Targeting

ADVANCED LAB: GENERATIVE AI

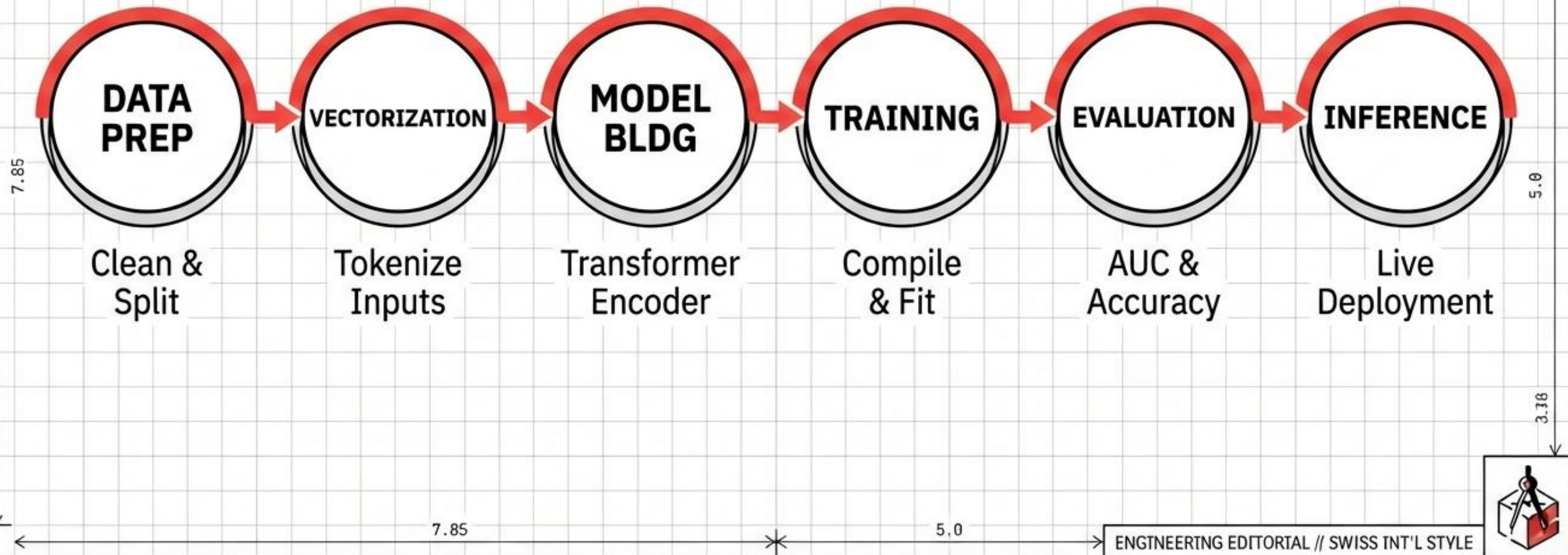


- **Creative** Output
- **Focus:** Text Generation
- **Output:** New Content (LLMs)
- **Use Case:** Content Creation

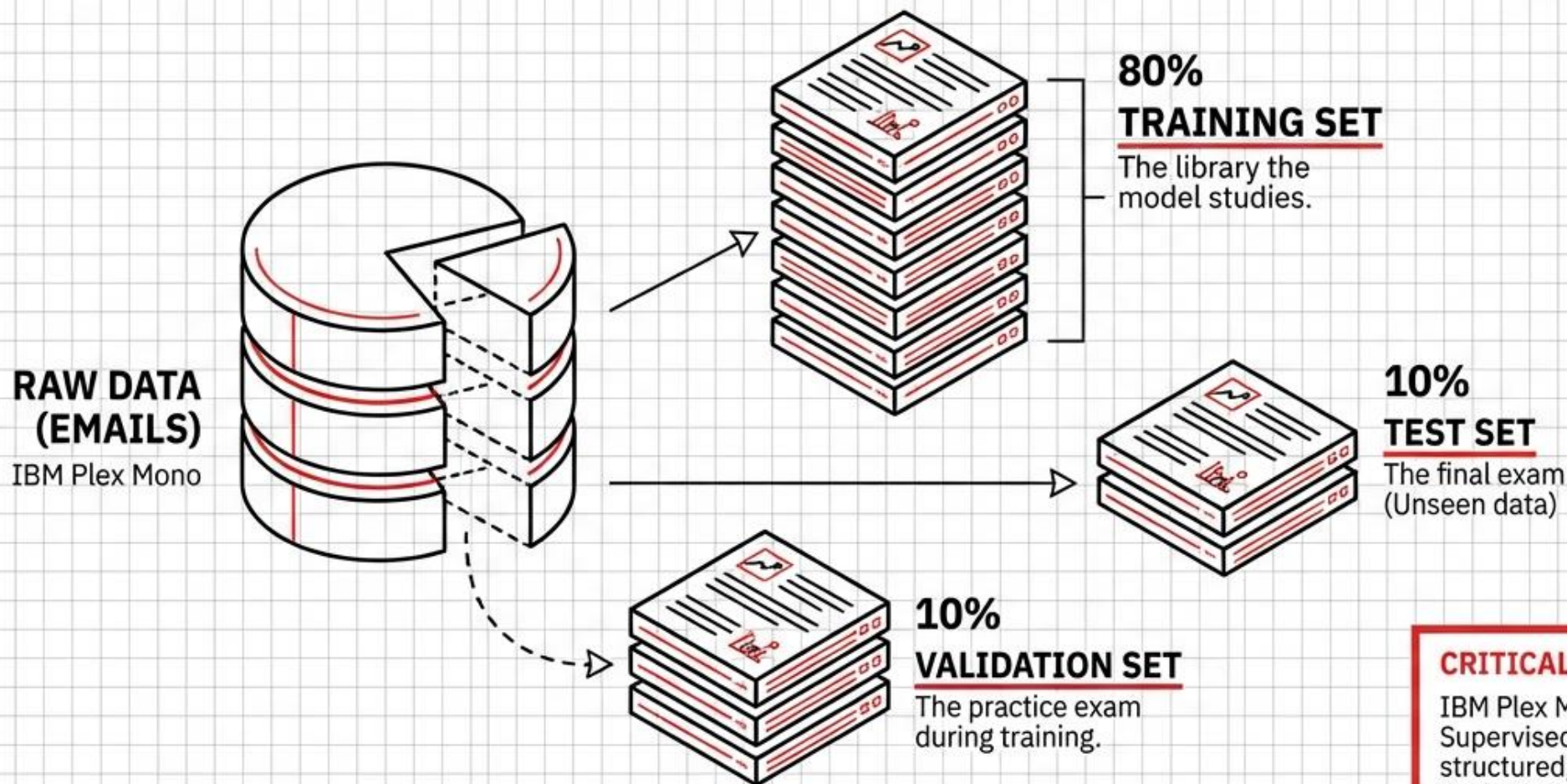
TAKEAWAY: WE MUST MASTER THE DETERMINISTIC PIPELINE BEFORE MOVING TO GENERATIVE MODELS.



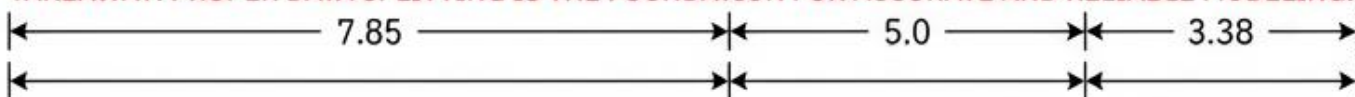
THE MACHINE LEARNING PIPELINE



STEP 1: DATA PREPARATION AND OPT MIZATION

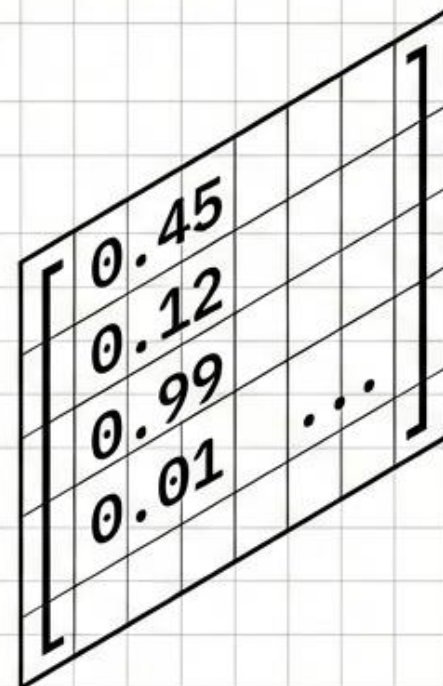
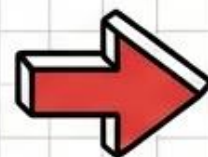
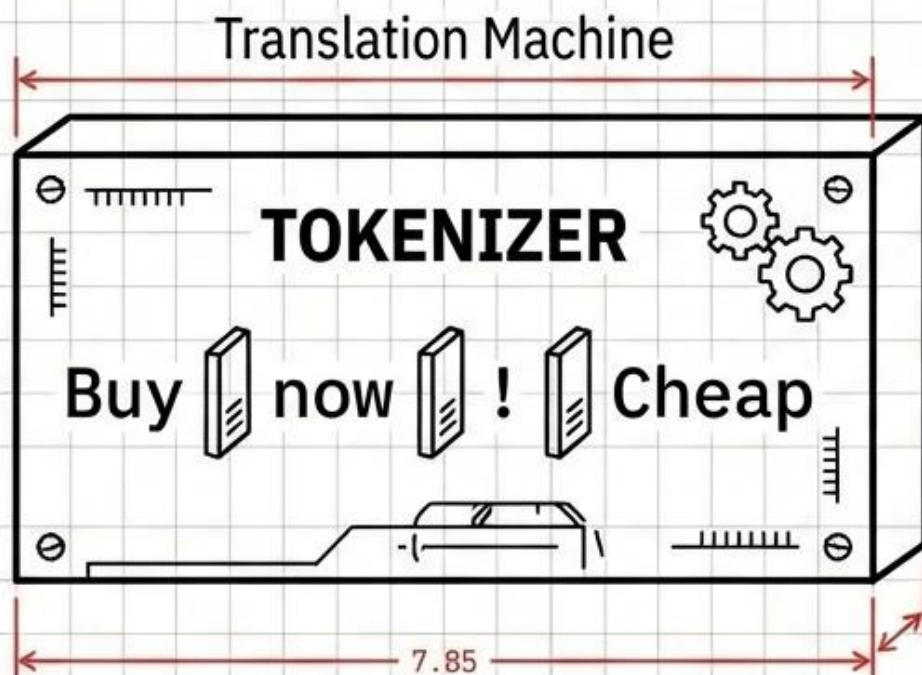
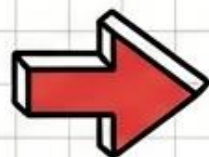


TAKEAWAY: PROPER DATA SPLITTING IS THE FOUNDATION FOR ACCURATE AND RELIABLE MODELING.



STEP 2: VECTORIZATION AND TOKENIZATION

Buy now!
Cheap
Rolex...



NUMERICAL VECTORS

Converting human language into machine mathematics.
The Transformer cannot read text; it only processes vectors.

7.85

5.0



STEP 3: BUILDING THE TRANSFORMER ENCODER

**Global Average Pooling
& Dense Classifier**

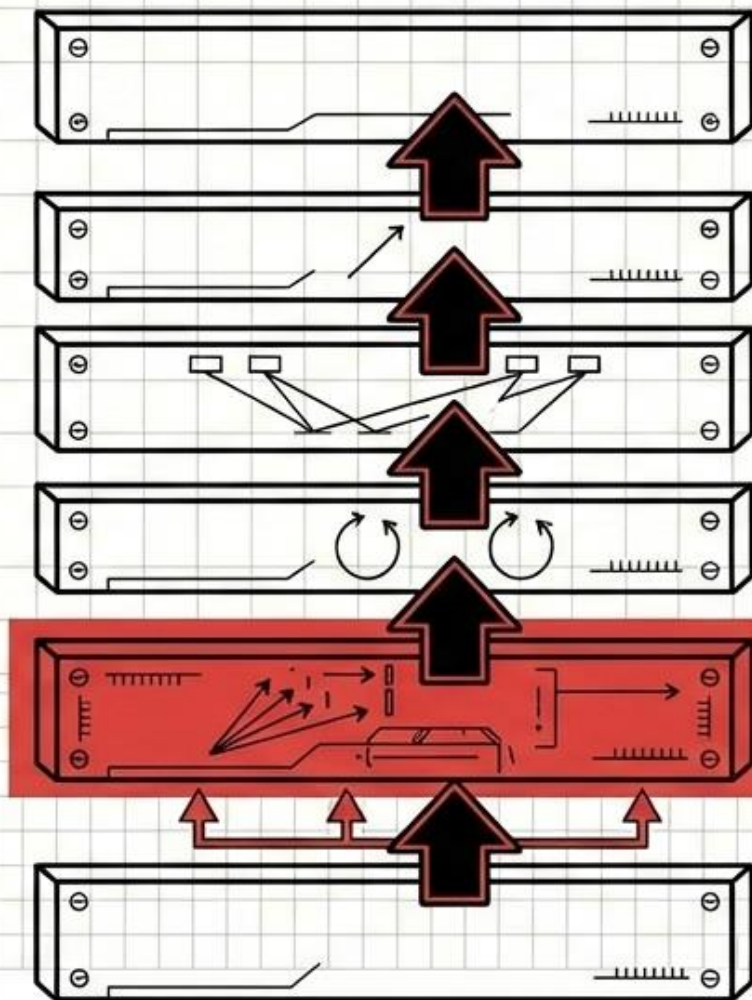
Add & Norm

Feed Forward

Add & Norm

**Multi-Head
Attention**

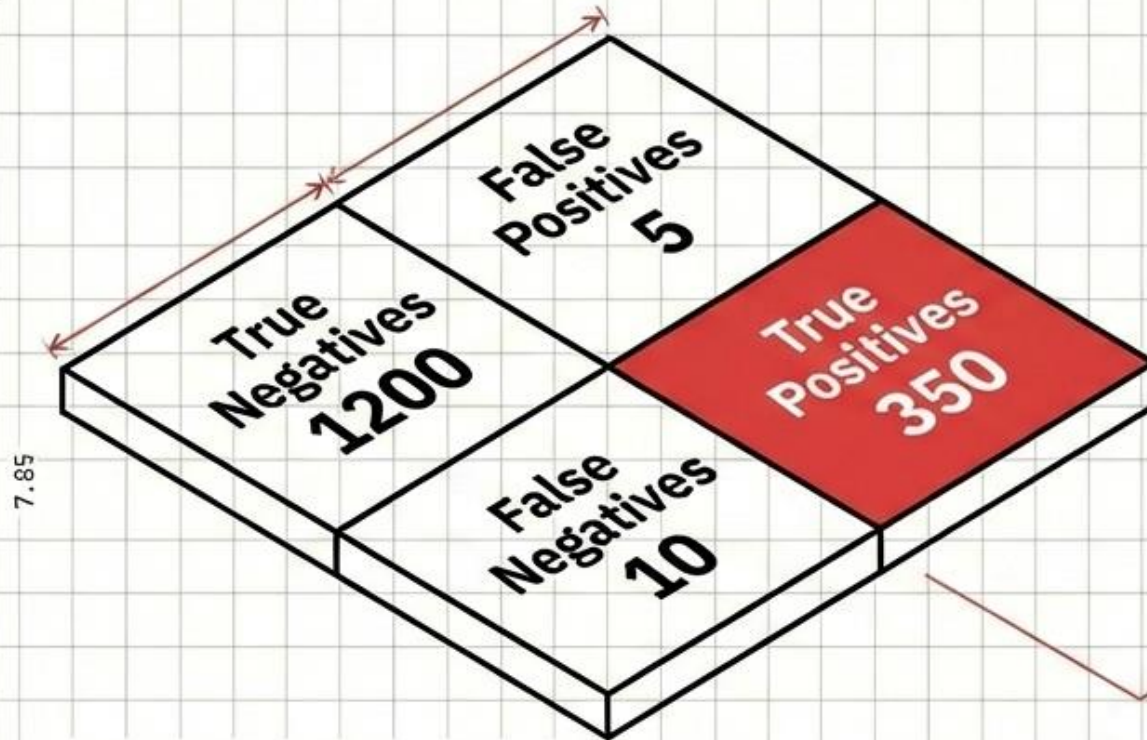
Input Embeddings



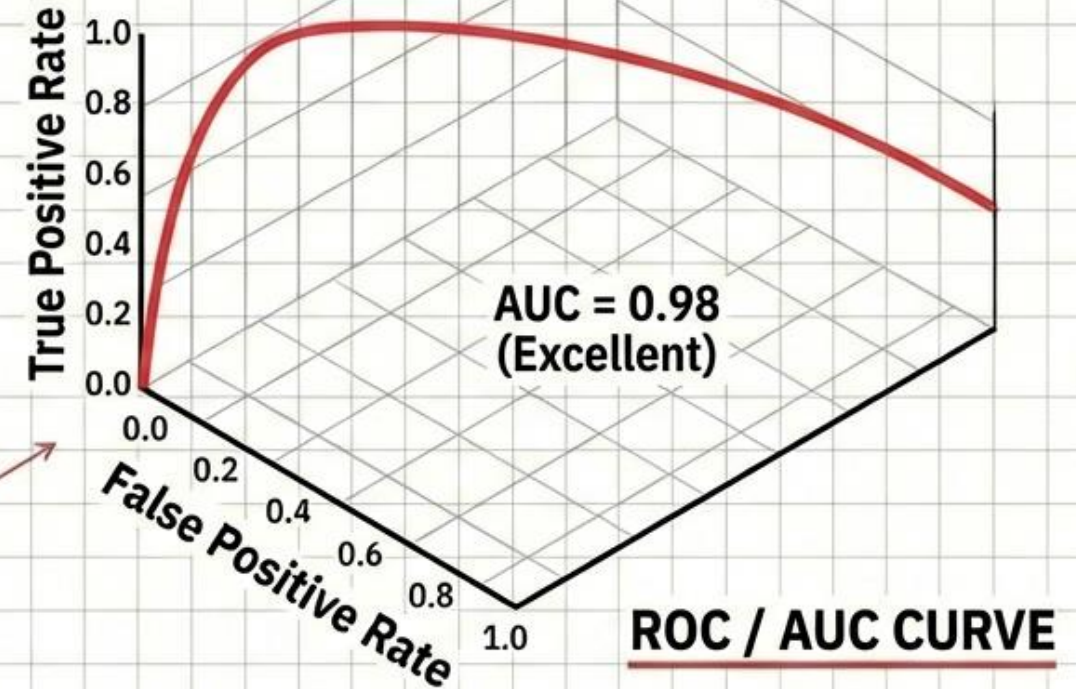
THE ENGINE

We are implementing a Deep Learning architecture. The Encoder understands context, while the Classifier makes the final binary decision.

STEP 4: TRAINING AND EVALUATION METRICS



CONFUSION MATRIX



ROC / AUC CURVE

THE GOAL: PROVING DETERMINISTIC RELIABILITY.

We validate against the Test Set to ensure the tool is ready for sales deployment.

**Receiver Operating Characteristic
Area Under Curve**



STEP 5: LIVE INFERENCE AND DEPLOYMENT

OPERATIONALIZING

1. Save trained artifacts.
2. Reload model.
3. Run real-time predictions on sales data.

Terminal

```
> LOAD_MODEL('spam_classifier_v1.h5')  
> MODEL LOADED SUCCESSFULLY.  
> INPUT: 'Urgent: Update your password  
now for free gift'  
> PROCESSING...  
> PREDICTION: [SPAM]  
> CONFIDENCE: 99.4%
```



CAPABILITY SUMMARY: FROM THEORY TO PROFICIENCY



1. DATA PREP

Can explain how text datasets are structured and split for ML.



2. TOKENIZATION

Can convert raw text into numerical vectors.



3. ARCHITECTURE

Can implement a Transformer Encoder block.



4. VALIDATION

Can evaluate performance using AUC and Accuracy metrics.

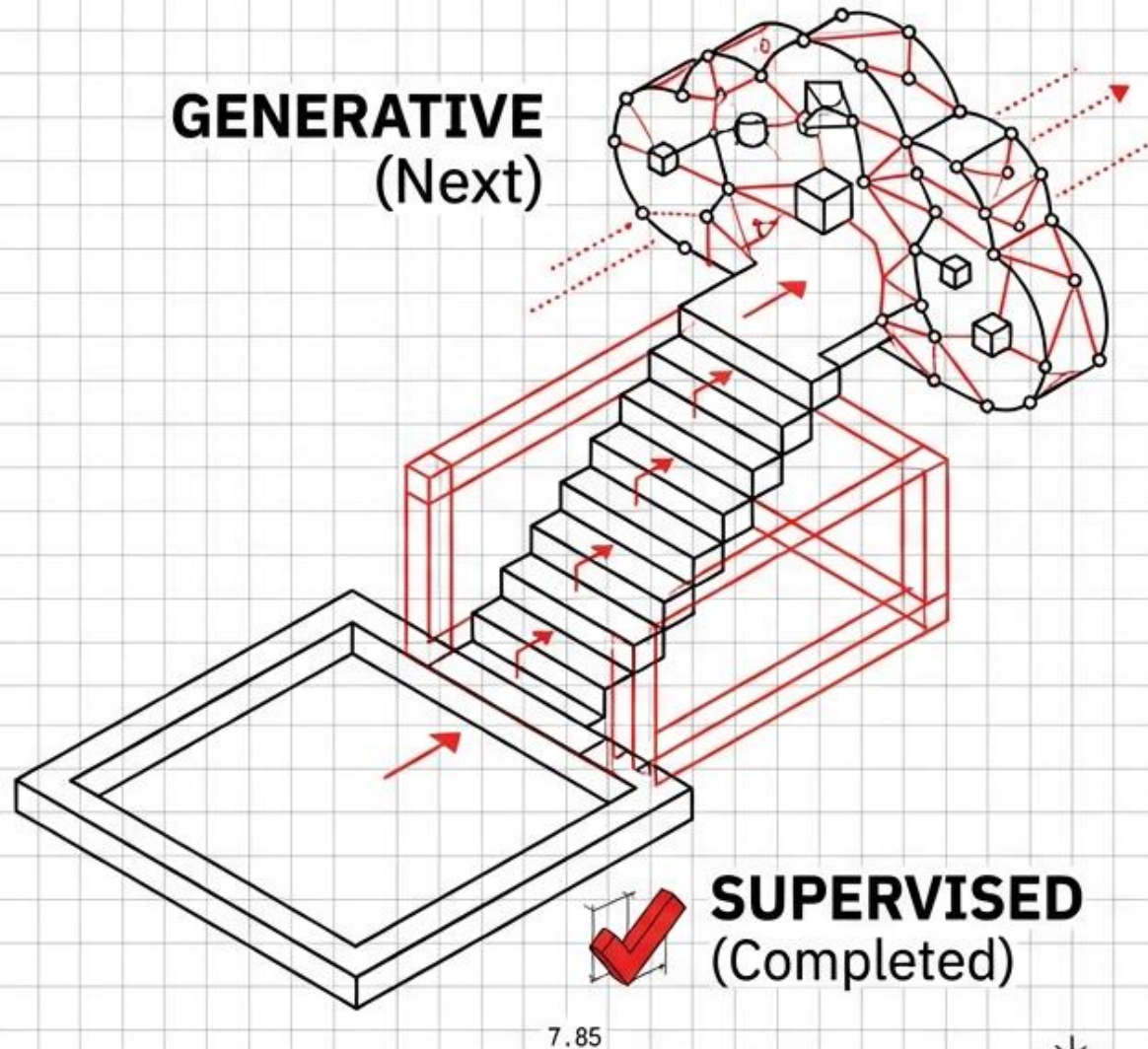


5. DEPLOYMENT

Ready to build internal predictive tools for 2026.



THE ROAD AHEAD: ADVANCED GENERATIVE AI



UPCOMING TOPICS:

- Large Language Models (LLMs)
- Generative Models (Creative vs. Deterministic)
- AI Red Teaming Training
- “AskTEC” Data Maturity Use Case



RESOURCES AND CALL TO ACTION

SOLUTION JUPYTER NOTEBOOK



[Internal Repository Link Placeholder]

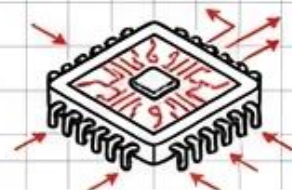
GOOGLE ML TRAININGS



Reference Materials

<https://fortinetcloudcse.github.io/genai-creator-lab/>

**“Equipping the team to help Fortinet
move faster through AI.”**



Q&A / APPENDIX

DEEPER DIVE

discussion points:

- The Math of Transformers (Attention Mechanisms)
- Specific Sales Targeting Use Cases
- Handling Data Imbalance

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

$$C = A \times B$$

$$\sum = \sum_{i=1}^n x_i$$

