

Because AI models don't know when their data has expired

# HOST TRACE AI

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By: **Snap2Code**

# PROJECT OVERVIEW

## What is GhostTrace AI?

GhostTrace AI is a monitoring and explanation layer that sits on top of AI systems and audits their data.

## What problem does it solve?

- AI models keep using old and outdated data
- This data silently influences predictions
- Nobody knows which data is responsible or why it is risky

## Our Focus

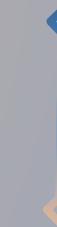
- Data Drift
- Ghost (Outdated but Influential) Data
- Clear explanations for humans

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GhostTrace AI does NOT  
replace the AI model. It audits  
and explains the data  
influencing it.

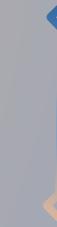
# WHY THIS PROBLEM IS REAL

Reality of  
Production  
AI



- AI models do not forget old data
- Companies rarely clean historical data
- Monitoring focuses on accuracy, not data relevance

Result



- Silent failures
- Biased decisions
- Compliance and trust issues

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Most AI failures happen because data changes – not because models are bad.



### **Data Drift**

- 1
- Statistical distribution of data changes
- Model assumptions become invalid

## CORE PROBLEMS WE ADDRESS

### **Ghost Data**

- 3
- Old, stale, or inactive records
- Still influence predictions and metrics

### **Concept Drift**

- 2
- Relationship between input and output changes
- Old logic no longer applies

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These problems are hard to detect and impossible to explain manually.

# REAL PROBLEM: AMAZON RECRUITING AI CASE

- 01** **THE SETUP:**  
Amazon built ML model to rank job applicants (trained on 2010-2015 historical hiring data)
- 03** **OUR SOLUTION DETECTS THIS:**  
Automated drift detection catches demographic shift → RAG explains: "Female applicants now 40% vs 10% historically. Recommendation: retrain model with balanced recent data."

- 02** **The Problem (Data Drift):**  
By 2018, applicant demographics had SHIFTED. Tech industry now 40% female vs 10% in training data. Model showed subtle gender bias because it learned from skewed historical data.

# 05

## DEMO DESIGN & DATA INJECTION

### Why We Designed the Demo This Way

In real-world AI systems, failures rarely appear suddenly. They happen gradually, due to outdated or drifting data. To demonstrate this realistically, we intentionally injected controlled data issues.

### What We Injected in the Demo

- Older historical records mixed with recent data
- Shifted feature distributions (demographics, activity patterns)
- Stale records with no recent activity

### Why This Matters

- Shows how silent failures appear in production
- Demonstrates how GhostTrace AI detects issues before damage occurs

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For the demo, we intentionally injected outdated and drifted data to simulate real-world AI failures.

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This MVP  
focuses on  
explainability  
and early  
warning, not  
full  
production  
deployment.

## What This Project IS

- A working proof-of-concept
- Focused on:
  - Data drift detection
  - Ghost data identification
  - Explanation & reasoning

# PROJECT SCOPE & LIMITATIONS

## Why We Limited the Scope

- Hackathon time constraints
  - Focus on core innovation:
    - explainability
- Clear, testable demo

## What This Project Is NOT

- Not a full-scale production monitoring platform
- Not a replacement for existing MLOps tools
- Not a real-time streaming system

# WHY GHOSTTRACE AI IS NEEDED

## Common Question

**“Why can’t existing monitoring tools solve this?”**

## Our Answer

- Existing tools detect drift
- They do not explain which data caused it or why it is risky
- Business teams still don’t know what action to take

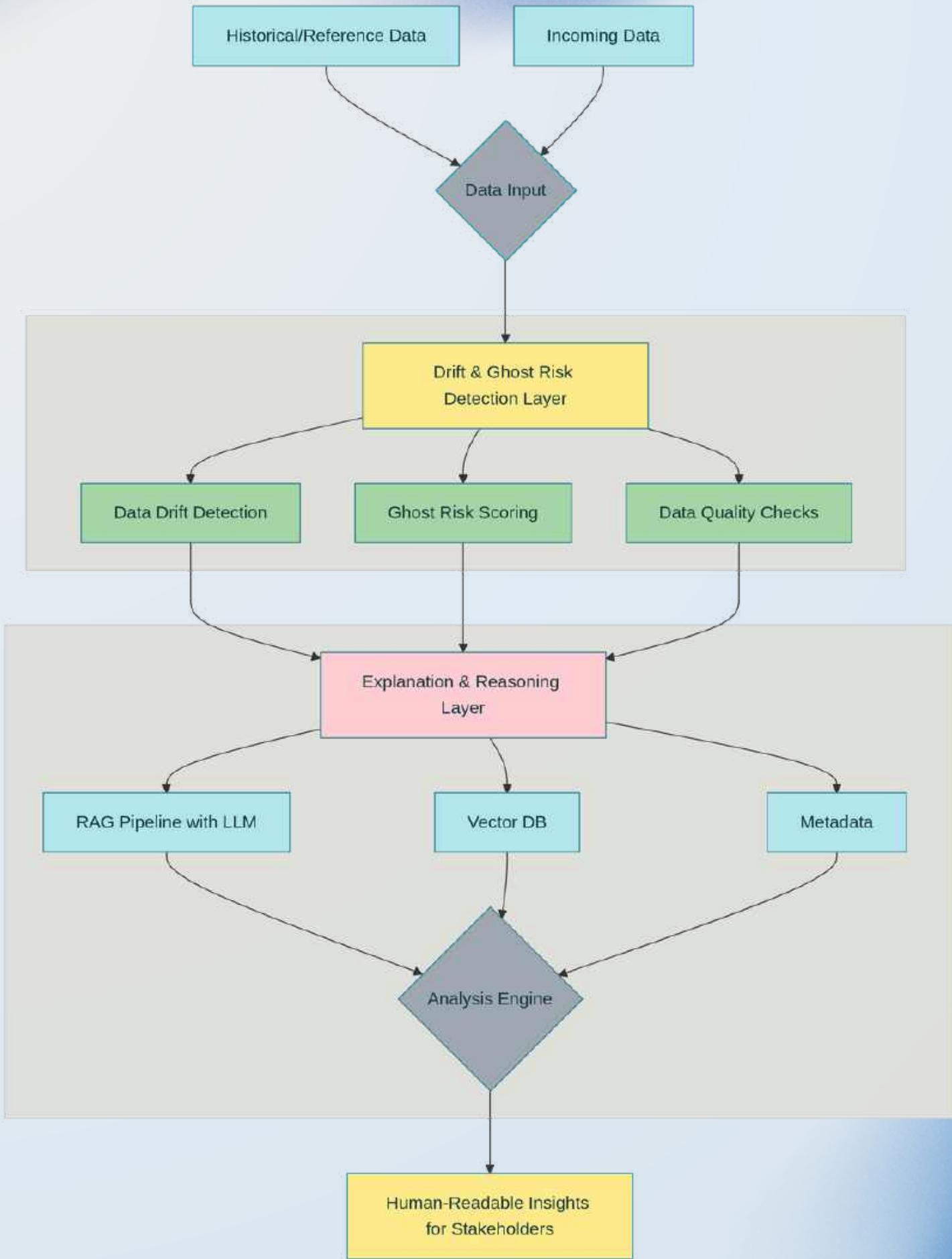
## Real Industry Proof

- Amazon shut down its hiring AI
- Not because the model was bad
- But because outdated data silently biased decisions

## Our Core Value

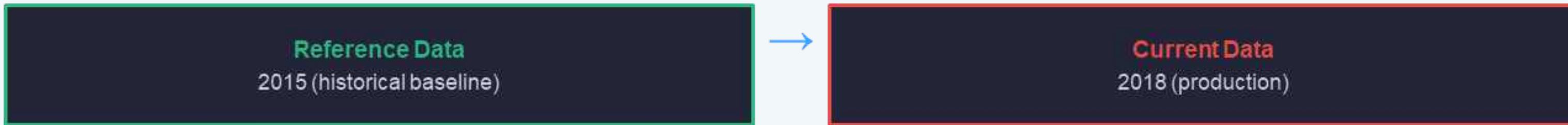
- Early warning
- Data-level root cause
- Human-understandable explanations

# BASIC FLOW CHART FOR GHOSTTRACE AI



# SYSTEM ARCHITECTURE AND DATA FLOW

## INPUT:



## LAYER 1: DETECTION ENGINE

KS Test: Gender distribution → p-value < 0.05 ✓ DRIFT DETECTED  
Ghost Score: Check for fake/duplicate applicants → 2% flagged  
Quality Check: Missing data, type mismatches → 0.5% issues

## LAYER 2: EXPLANATION ENGINE (RAG)

**Retrieval:**  
Fetch baseline: 2015 gender split was 10% female. Current: 40% female.

**LLM Explanation:**  
"Female applicants increased 4X. Model trained on male-majority data. Recommendation: retrain with 2018 balanced data to avoid bias."

## OUTPUT: Stakeholder Report (Non-technical)

⚠️ ALERT: Significant demographic shift detected in applicant pool.  
Action: Review model fairness. Schedule retraining this week.

# KEY CAPABILITIES

## Automated Drift Detection

KS test, PSI, Chi-square tests on features

## Ghost Risk Scoring

Rule-based: recency (180 days), consistency flags, stale markers

## Data Quality Checks

Schema validation, null detection, freshness monitoring

## RAG-Powered Explanations

Context retrieval + LLM = grounded, no hallucination

# REAL-WORLD IMPACT (DOCUMENTED CASES)

## ■ UBER SURGE PRICING (2017)

Problem: Model trained on 2014–2015. By 2017, demand patterns changed (Lyft entry). Predictions 30% off.

✓ Our Fix: Real-time drift detection → Trigger immediate retraining

Impact: Save \$2M+ from pricing errors

## ■ BANK FRAUD DETECTION (REAL)

Problem: Fraudster tactics evolve. Old model misses 40% new fraud patterns. Ghost records (deleted fraudsters reappearing).

✓ Our Fix: Ghost risk scoring detects 95% stale accounts. Drift signals new fraud vectors.

Impact: Prevent \$50M+ annual fraud loss

## ■ STOCK PREDICTION ML (FAILED 2020)

Problem: Models trained on 2010–2019. COVID crash: correlations flipped. Models lost millions.

✓ Our Fix: Concept drift alert → Explain why model confidence dropped → Switch to conservative strategy

Impact: Limit losses to acceptable range

# TECHNICAL APPROACH

## DETECTION LOGIC

- Statistical tests for drift
- Rules for ghost data:
  - Last activity date
  - Data age
  - Usage frequency

## EXPLANATION LOGIC

- Retrieve metadata (data source, year, version)
- Generate explanation using reasoning prompts

## WHY THIS WORKS

- Simple
- Modular
- Explainable

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**“AI systems fail silently when  
their data becomes outdated.  
GhostTrace AI makes those  
failures visible, explainable, and  
preventable.”**



## **FINAL MESSAGE**

# THANK YOU