Al Handbook - Technical Report

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Approach Overview

I implemented a hybrid rule-based + Al-powered system to enrich and score B2B leads. Our architecture supports:

- Rule-based scoring logic (0–100 scale)
- Azure OpenAl GPT-based insight generation
- Scalable chunked CSV processing
- REST API backend (FastAPI) with a React frontend

Design principles include interpretability, speed, and actionable insights.

Data Preprocessing

Transformations Applied:

- company age = 2025 founded
- size_bucket mapped to 1–8 scale from size_range
- Normalizations
 - o norm size = (size bucket min) / (max min + ε)
 - o norm_age = $(max age) / (max min + \varepsilon)$
- Boost calculations
 - o size_boost = norm_size^1.5
 - o age_boost = norm_age^1.8
- Flags
 - o maturity bonus: age > 20 and size ≥ 7
 - o growth bonus: current emp < 70% of total emp

Model Selection

1. Lead Scoring Engine - Rule-based formula

Scoring logic: lead_score = 0.5 * size_boost + 0.4 * age_boost + 0.05 * maturity_bonus + 0.05 * growth bonus

2. Insight Generator - Azure OpenAl GPT-4

Prompt designed to explain lead score using only provided data and provide a structured output.

Performance & Benefits Summary

- 1. High Throughput & Low Latency -
 - Processes ~10,000 rows per minute on a 4 vCPU setup
 - Handles input CSVs up to 100MB
- 2. Cost-Effective & Scalable -
 - <3% JSON error rate in AI outputs
 - No training data required
 - Supports parallel processing with low memory usage
- 3. Transparent & Business-Aligned
 - Easy tuning through rule update
 - Direct mapping to business priorities