Al Maturity Model for Production (AlM2prod)

Enterprise Architecture Al Tutorial Draft June 15, 2023 John R. Frank, PhD

Al Maturity Levels for Production Software Engineering

- L5 User experience changes inspired by AI process.
- L4 Using bake-offs and error analysis to improve user effectiveness.
- L3 Capturing judgements on real test data with written guidelines.
- L2 Recording & studying metrics of usage and user feedback (stories).
- L1 Identified outputs that impact human effectiveness & that can be judged.
- L0 Documented APIs, versioned, regression tests, used by other engineers.

Foundational AI Strategies for Software Engineering



Three Canonical Examples

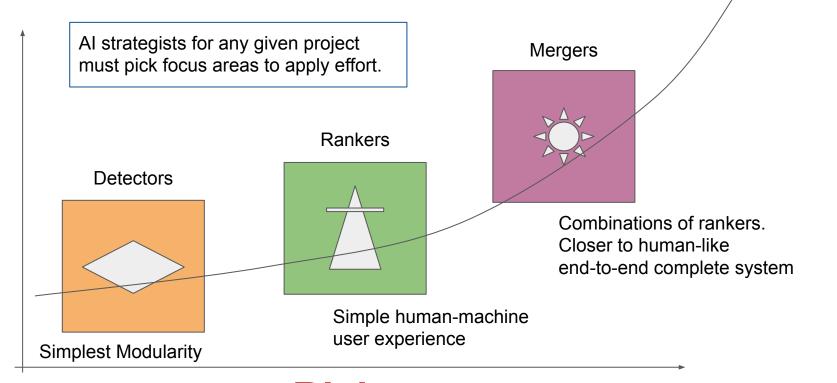
Al Engineering Strategy Template (best practices)

- Measure Quality: Identify restricted Turing tests at module boundaries.
 - 1.1. General Turing tests: open-ended conversation \rightarrow single judgement of is intelligent or not.
 - 1.2. *Restricted* Turing tests: **simplified story templates** → repeated judgements that we **count**.
 - 1.3. Leverage standard metrics, e.g. from NIST

- 2. **Use Real Data:** Establish baseline measurements on operational data.
 - 2.1. Use existing operational system or simplest possible rule-based function.
 - 2.2. Make a test harness that records human judgements usually new code.
 - 2.3. Crucial: document guidelines with examples of edge cases, so we can train more judges.

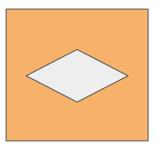
- 3. **Improve Quality:** Run **offline** contests or "bake offs":
 - 3.1. Periodically upgrade input/output APIs of modules, so new AI approaches can run on tests.
 - 3.2. Crucial: save historical outputs of test runs for each iteration of new Al algorithms.
 - 3.3. Error Analysis: periodically study errors in past runs to identify opportunities for paths forward.

Reward



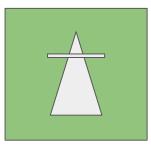
Risk

Three Standard Templates for AI Engineering



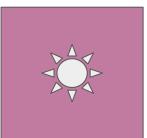
Detectors:

- Input: small complex data + defined type of thing, output is trivial yes/no (often plus/minus)
- Restricted Turing Test: human can read/see/hear input to determine yes/no.
- Examples:
 - Is this a phone number from region X?
 - Is this a photo of thing X?



Rankers:

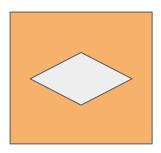
- Input: massive complex data + user request, output is ordered list of responses.
- Restricted Turing Test: human reviews list, accepting/rejecting answers.
- Examples:
 - What should I read about topic X?
 - Give me new factoids of type X to add to my report on topic Z.



Mergers:

- Input: massive complex data + user request + schema of ideal output → populated schema
- **Restricted Turing Test:** human accepts/rejects data in each field.
- Examples:
 - Route me to points A, B, C efficiently given current traffic.
 - Organize my chemistry experiments to find best procedure for synthesizing X.

Metrics for Detectors



Detectors:

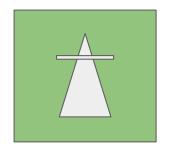
- Several standard metrics computed from building a "confusion matrix"
- Different metrics aim at different goals
- Often implemented with a numerical score computed for each input and then a threshold that decides whether an output is positive (above threshold) or negative (below threshold).

Confusion Matrix (example)		Human Judgement		
		True	False	Totals
System Output	Positive	9,182	295	9,477 (above threshold)
	Negative	6,485	48,279	54,764 (below threshold)

Example:

- Rate of false positives
 3% = 295 / 9477 → Precision = 97%
- Rate of missed detections
 88% = 48279 / 54764 → Recall = 12%
- Aggregate "accuracy" is:
 24% = (9,182 + 6,485) / (9,477 + 54,764)
- F1 score
 harmonic mean of precision & recall:
 2 * prec * recall / (precision + recall)
 In this example: 21%

Metrics for Rankers



Rankers:

- Metrics count effort required by user to read down the list and find good result(s)
- Nuanced judgment about what counts as a "good" result, e.g. not redundant with earlier hits.
- Like a detector with threshold lowered to allow higher recall with lower precision.
- Expected Reciprocal Rank (ERR) is a widely used scoring function.

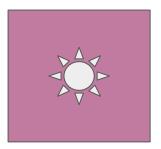
Q1	Q2	Q3
R1.1 X R1.2 X R1.3 X R1.4 X R1.5 X	R2.1 X R2.2 ✓ R2.3 ✓ R2.4 ✓ R2.5 ✓	R3.1 X R3.2 X R3.3 X R3.4 ✓ R3.5 ✓

Ranking metrics tend to be precision focused.

Quantifying recall requires more judgments.

1 2 4
$$\implies$$
 ERR = avg(1 + $\frac{1}{2}$ + $\frac{1}{4}$) = 7/12 first good hit

Metrics for Mergers



Mergers:

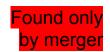
- Metrics based on comparison of output with a human-assembled work product.
- Detector-like: Count human accept/reject elements from merger?
- Quantify Recall: What did the human find without the merger, what did the merger add?
- Gap analysis → Rumsfeld's Matrix of Epistemic Uncertainty

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Known Knowns	Known Unknowns	
Unknown Knowns	Unknown Unknowns	



Found only by human

