

How do you *know* if a skill covers what you need?

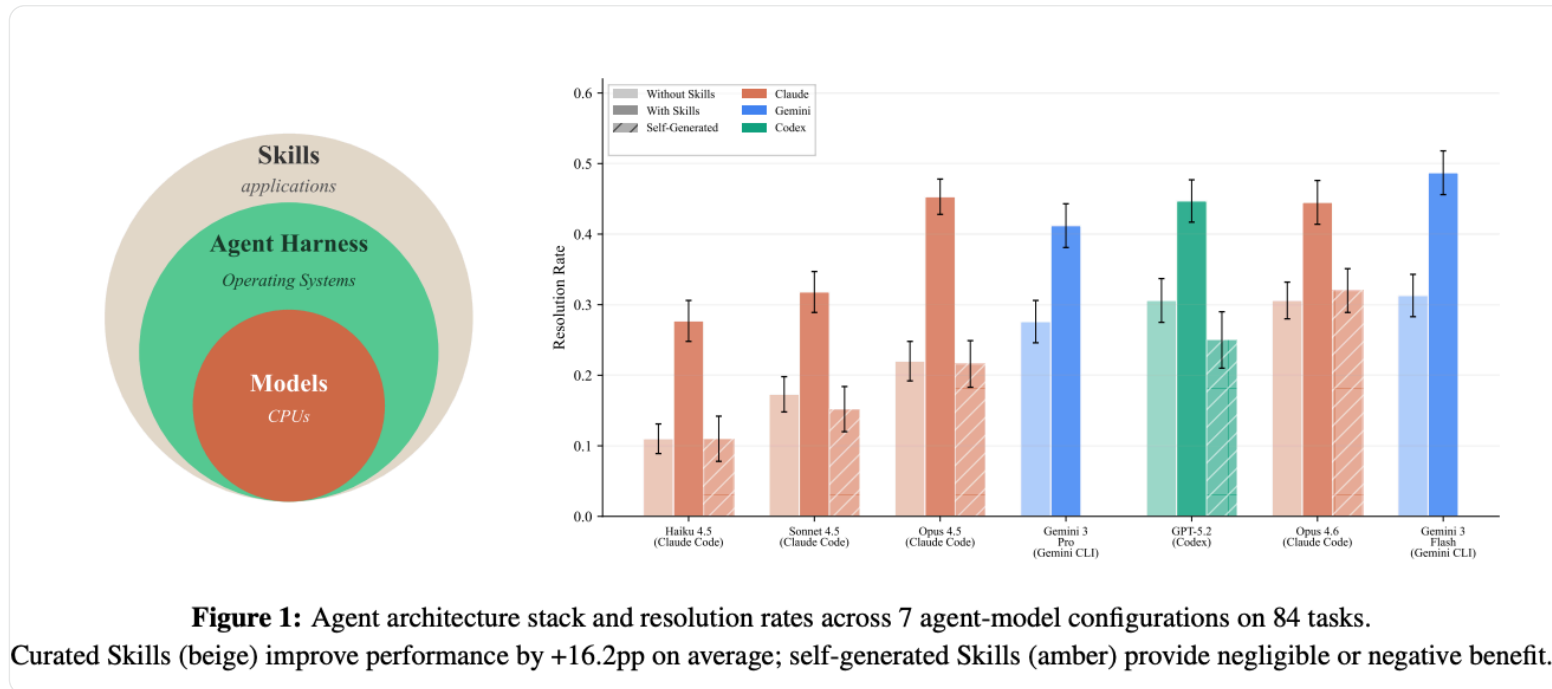
AI skills are curated knowledge bases that teach an assistant how to work in a specific domain — a codebase, a house style guide, a field of research, a body of legal precedent. Skill Cards audits them against the reference material they claim to cover.

Sruly Rosenblat · AI Disclosures Project



A sustainable marketplace needs a way to tell the *good* from the *bad*.

Skill quality differences are real and measurable.



In the study shown above, self-generated skills performed significantly worse than human-written ones. Browsing a marketplace doesn't show you any of that. arxiv.org/pdf/2602.12670

Skills on the same subject aren't *interchangeable*, but they are hard to *distinguish*.

SKILL.MD

fastapi-expert

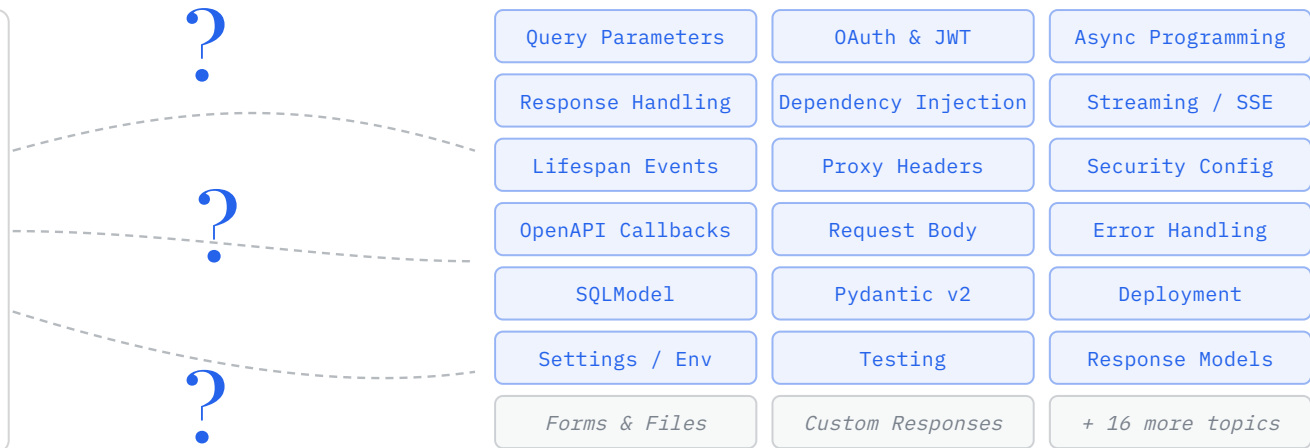
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name: fastapi-expert  
description: Use  
when building async  
Python APIs with FastAPI  
and Pydantic V2.  
license: MIT  
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FastAPI Expert

Deep expertise in async Python, Pydantic V2, and production-grade APIs.

5,453 words

DOCS/FASTAPI/ · 166,616 WORDS · 37 TOPICS



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Query Parameters	OAuth & JWT	Async Programming
Response Handling	Dependency Injection	Streaming / SSE
Lifespan Events	Proxy Headers	Security Config
OpenAPI Callbacks	Request Body	Error Handling
SQLModel	Pydantic v2	Deployment
Settings / Env	Testing	Response Models
Forms & Files	Custom Responses	+ 16 more topics

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Does the left actually *cover* the right?

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There are dozens of skills for each topic. Nobody can check them all.

See the differences clearly without reading every skill.

FastAPI docs · 166,616 words · 37 topics, measured against three skills from the Claude skill marketplace.

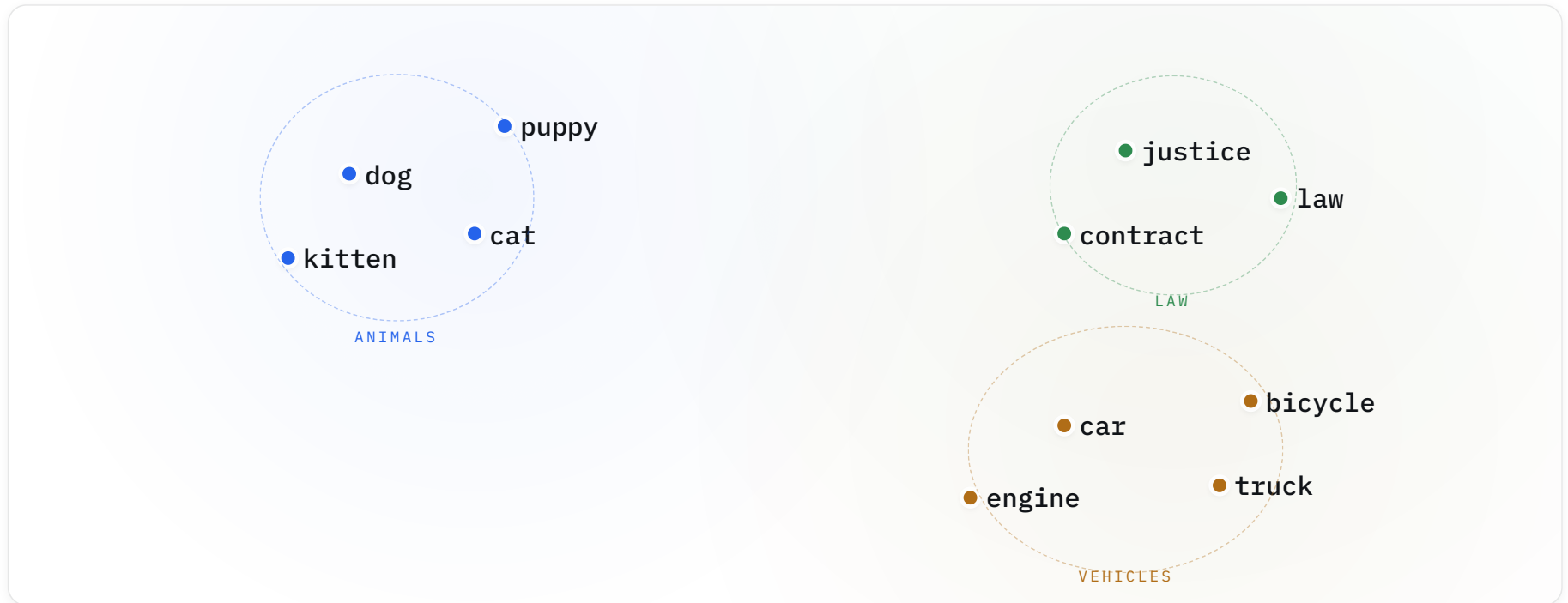
focus tier | ● Heavy ● Moderate ● Light ● Trace ● Top focus per topic

TOPIC CLUSTER	FASTAPI-EXPERT 5,453 words	FASTAPI-PRO 901 words	FASTAPI-TEMPLATES 1,402 words
SQLModel Database Integration	35.7%	0%	39.1%
OAuth2 and JWT Authentication	23.6%	0%	28.3%
FastAPI Introduction and Basics	6.0%	27.8%	4.3%
Deployment and HTTPS Concepts	2.2%	16.7%	0%
FastAPI Security and Configuration	7.7%	0%	15.2%
Contributing and Community Support	0%	16.7%	0%
FastAPI Response Handling	0%	0%	0%

Language models can lay out text in a space where similar ideas end up *next to each other*.

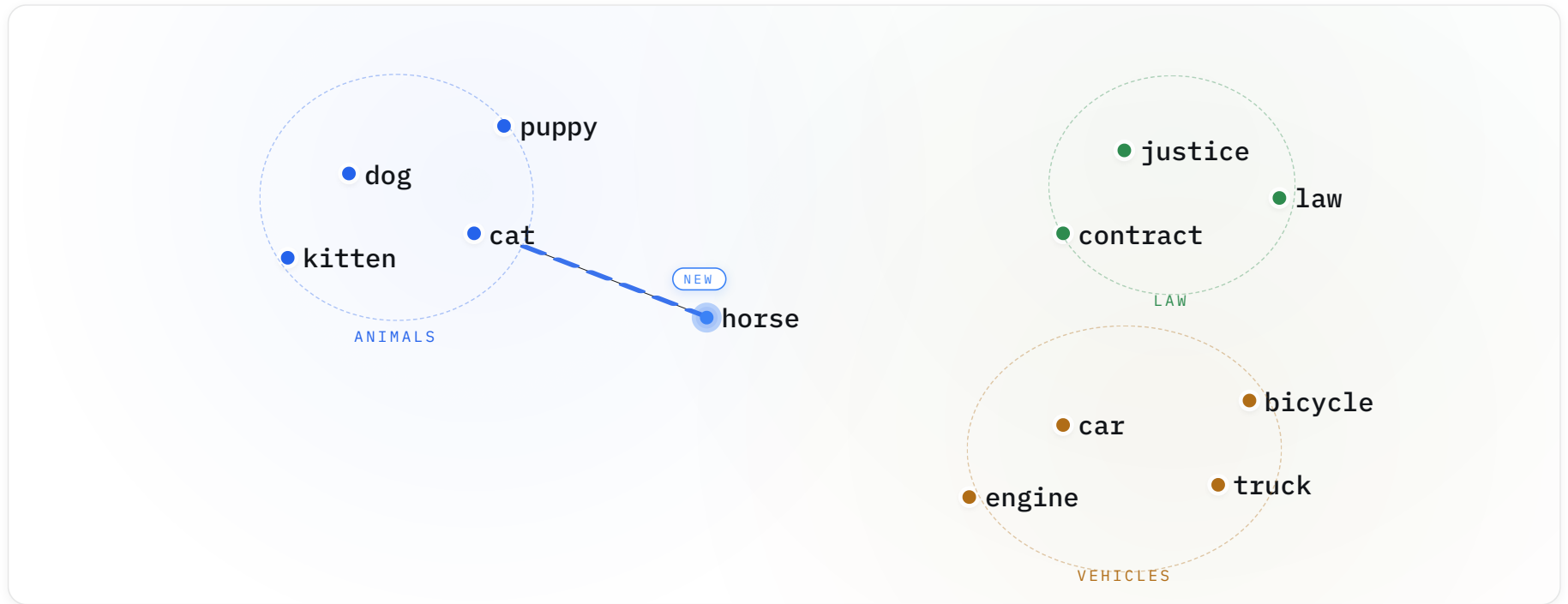


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Dots that sit close together mean similar things.

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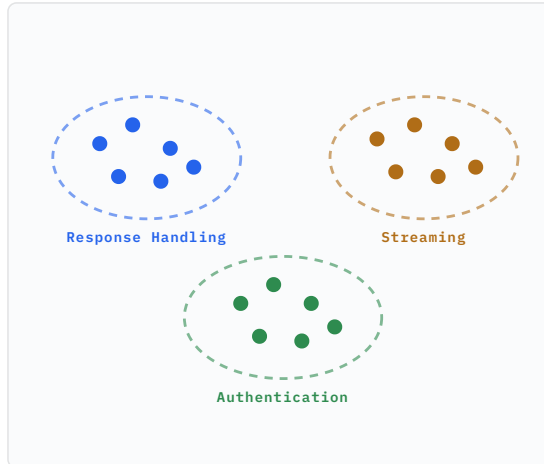
How to compare skills to reference material.

1
Break the docs into pieces.



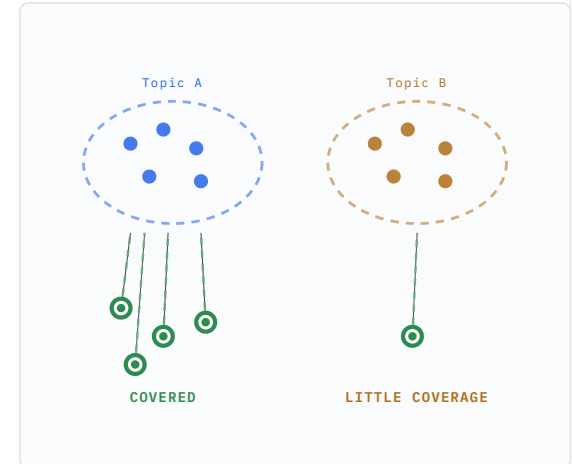
Each piece becomes a point in the meaning-space.

2
Cluster, then label.



Nearby points form topics. An LLM reads each cluster and gives it a name.

3
Measure the skill against each topic.



If a skill chunk lands inside a topic, it's **covered**. If nothing lands, it's a **gap**.

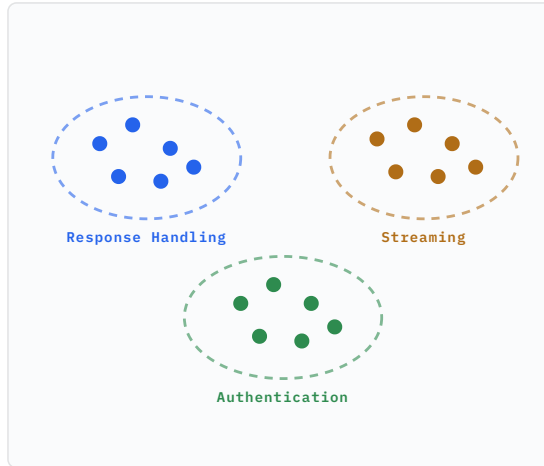
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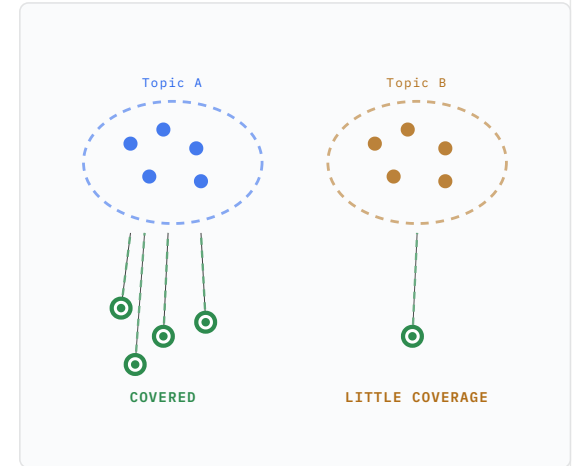
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The skill goes through the same mapping process. Coverage is just geometry after that.

A coverage score is not a quality score.

MISSES Whether the skill is **correct**.

MISSES Whether the skill is **well-written** or usable.

MISSES Whether a topic is **important** or niche.

FINDS Topics the skill **never addresses**.

FINDS Two skills that cover the **same topics**.

FINDS What topics the skill **misses**.

FINDS Where a skill puts its **focus**.

FINDS How a skill **compares** to others.

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*Finds what's **missing**. Can't judge what's **there**.*

The same system works for anything made of text.

Code docs were one example. The math doesn't know what it's measuring – two piles of text, one mapping, one coverage score.

a technical textbook ↔ a coding skill *Which chapters' techniques does the skill actually know?*

clinical guidelines ↔ a medical skill *Which conditions does the skill leave out?*

case law ↔ a legal skill *Which precedents does the skill reference when drafting?*

a research corpus ↔ a literature-review skill *Which subfields did the skill's review skip?*

a brand style guide ↔ a copywriting skill *Does the skill stay on-brand across every voice it writes in?*

your reference material ↔ your skill *The same audit runs against whatever text you're writing against.*

Available tools

Read-only MCP server at `skill-cards-mcp.fly.dev/mcp/`.

<code>list_doc_sets</code>	Every doc corpus the server knows.	<code>list_skills</code>	All skills, or skills for one corpus.
<code>get_skill</code>	Shape of one skill (per-file chunk counts).	<code>get_doc_topics</code>	Topic clusters in one corpus.
<code>get_skill_coverage</code>	Per-topic coverage for one skill.	<code>compare_skills</code>	Side-by-side coverage for N skills.
<code>find_coverage_gaps</code>	Weakest topics for one skill.	<code>get_topic_detail</code>	Drill into one cluster.
<code>search_skill_chunks</code>	Lexical search over a skill.	<code>search_doc_chunks</code>	Lexical search over a corpus.

any agent → `skill-cards-mcp.fly.dev/mcp/` → coverage JSON

Thanks.



SCAN FOR THE LIVE DASHBOARD

skill-cards-mcp.fly.dev

MCP endpoint at skill-cards-mcp.fly.dev/mcp/

LIVE DASHBOARD + MCP

skill-cards-mcp.fly.dev

SOURCE CODE

github.com/SrulyRosenblat/skill-cards

CONTACT

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