

わたしはだれでしょう?

- Director-NextGen Solutions, Planit (an NRI Company)
- Distinguished Speaker by ACM
- FOSS Fan
 - Project Leadership & Committer Selenium.dev
 - Software Freedom Conservancy
 - o Committer Appium.io
 - K8s-Selenium
 - o O11Y-Selenium
- DevOps & Accessibility Advocate





Japan's AI Demand Will Increase 320x by 2030, Industry Leader Says at NVIDIA AI Day Tokyo

At the event, which highlighted sovereign AI advancements in Japan, Kuniyoshi Suzuki of SoftBank Corp. discussed foundational technologies needed for AI expansion across the nation.

October 1, 2025 by Masataka Osak



Japan will see a 320x increase from 2020 in demand for Al computing power by 2030.

Kuniyoshi Suzuki

Senior Director, Cloud AI Service Division SoftBank Corp.

ARTIFICIAL INTELLIGENCE

Japan AI developers see path to success in localized training data

SoftBank developing large language model based on 9,000 hours of conversations

NRI: Barriers to overcome for AI adoption

Three Barriers That Must Be Overcome for AI to Spread Further

Barrier No. 1:

Accessibility

- Is it easy for anyone to use?
 - A certain level of Al literacy is required, and so mastering its use is difficult
 - However, in terms of using Al (in particular, generative Al), it can be utilized in a similar manner to search functions

Barrier No. 2:

Transformation of Values

- Will the use of Al significantly change traditional values?
 - Utilizing AI in situations normally handled by humans will necessitate big changes

Barrier No. 3:

Can the Surrounding Environment Keep Up?

- Can it exhibit network effects?
 - lt must go beyond the community level to achieve functionality improvements

Where We Are:

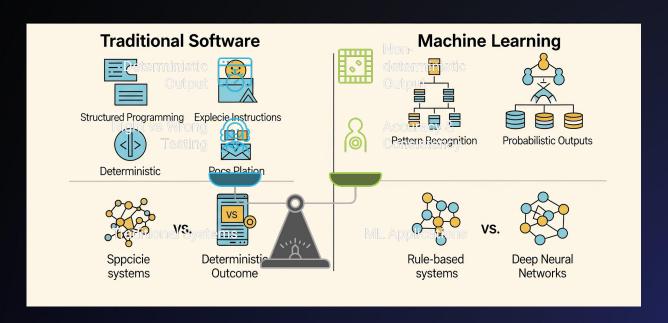
Tens of millions of people in Japan already have experience using Al

In the era of AI, it's not about machines replacing humans—it's about humans leveraging AI to unlock new potentials. In test automation, AI isn't just a tool; it's a partner in driving innovation and excellence.

3

Software Paradigms: Traditional vs. Machine Learning







The Testing Reality Check

The Current State of Testing in Al



Traditional Automation

Limited autonomous task performance

The Reality Gap

AI- Assisted Testing

Independent task execution achieved

The Vision....

Suggest Test Cases

Al proposes new test cases based on production patterns.

Adapt to UI Changes

Al automatically adjusts to UI <u>updates.</u>



Identify Risks

Al detects potential risks not initially considered.

Discover New Flow

Al identifies new user flows during testing.



Let's talk Al...



Artificial Intelligence

A piece of technology that can perform tasks that mimic human intelligence, such as those requiring reasoning, problem-solving, decision-making, or language understanding.





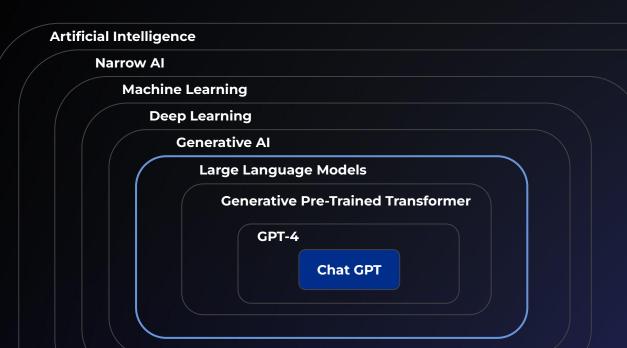
Taxonomy of Artificial Intelligence





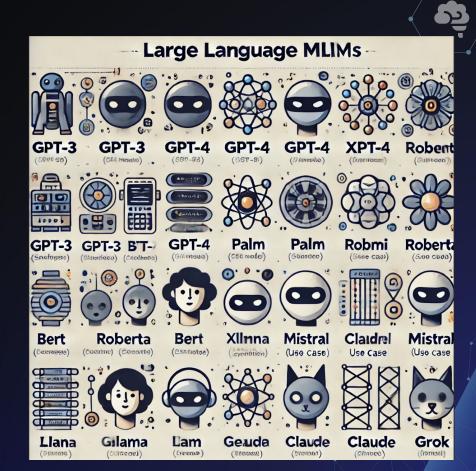
LLM's





Large Language Models

- ★ GPT-3 OpenAl
- ★ GPT-4 OpenAl
- ★ BERT Google
- ★ T5 Google
- ★ PaLM Google
- ★ RoBERTa Facebook Al
- ★ XLNet Google/CMU
- ★ LLaMA Meta AI (Facebook)
- ★ Gemini Google
- ★ Mistral Mistral Al
- ★ Claude Anthropic
- ★ Grok xAI (Elon Musk)





Al Agents

"Al Agents are the shift from asking 'Can you do this task?' to saying 'You handle this domain.'"

"They're specialized AI systems that can act independently within their area of expertise - but each works alone."



Agentic Al

"Agentic AI is the shift from asking 'What should I do?' to declaring 'Here's what I want to achieve.'"

"It's AI that doesn't just respond - it takes initiative, learns from experience, and persistently works toward your goals."



Al Agents Vs Agentic Al



Agentic AI is like the concept of an autonomous vehicle

An AI agent is like a Tesla on Full Self Driving mode

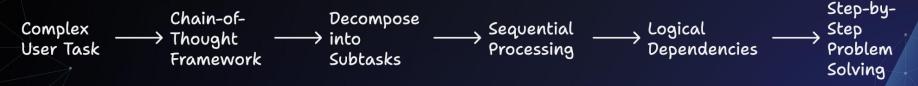
How Agentic AI works... Task Decomposition



Chain-of-Thought (CoT) serves as the primary decomposition framework

Complex user tasks are systematically broken down into manageable subtasks

Sequential processing enables step-by-step problem solving



Made with > Napkin

Chain of Thought

Standard Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The answer is 27.



Chain-of-Thought Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls. 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they have 3 + 6 = 9. The answer is 9.

How Agentic Al works... Adaptive Learning



Plan : Strategic task analysis and resource allocation

Execute: Active implementation of planned actions

Reflect: Performance evaluation and outcome assessment

Refine: Strategy optimization based on reflective insights



"Smart agents are great, but what if they have no idea what they're talking about?

That's where retrieval-augmented generation — or RAG — changes the game."

Knowledge Enhanced LLMs



Incorporating knowledge into LLMs can empower their memorization and reasoning abilities

Retrieval Augmented Generation

RAG Vs Agentic RAG



Static Info retrieval with No decision making

Single Step queries and response

Single LLM for all the task

No human intervention: Input In & Output

Agentic RAG

Autonomously decide what to retrieve and how to act on it

Multi-step reasoning and actions with multi-agents

Right LLM for right task

Human in the loop as necessary





Agentic Al workflow

Agentic Al Spectrum





Gen Al: Prompt Chaining

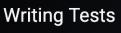
Agentic Workflow

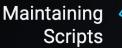
Semi-Autonomou s Agents Fully-Autonomous Agents

Where are we now...



Agentic AI in Quality





Reactive Solution Debugging

Isolated Testing



Traditional Testing



Defining Goals



Guiding Agents



Proactive Risk Management



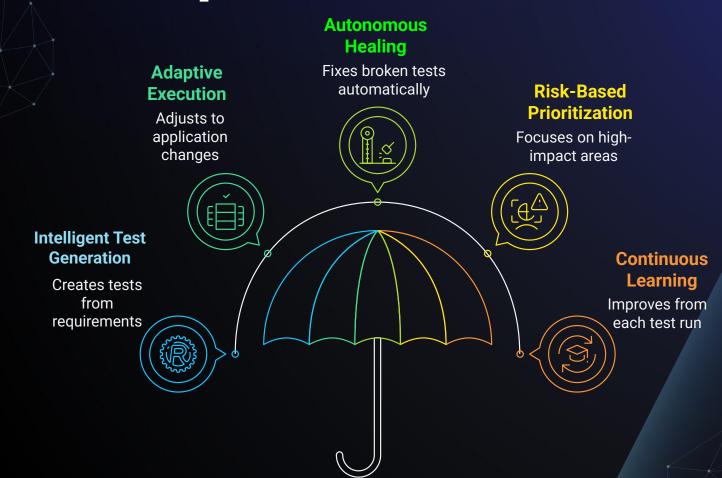
Collaborative Intelligence





Agentic Al Capabilities...

manojkk.com



ණි



What Powers Agentic Al?

Agents Together...

Orchestrator Agent

Manages overall testing strategy and workflow

Learning Agent

Continuously improves from testing results

Specialist Agents

Focus on specific testing domains

Communication Layer

Enables agent collaboration and data exchange







"Let's Build One Together"



Scenario

User Story to BDD Gherkin format





UserStory to BDD



UserStory Analyser

Agent responsible for analyzing user stories.



TestScenario

Generator

Agent that generates test scenarios.



BDD

Generator

Agent that creates BDD scenarios.

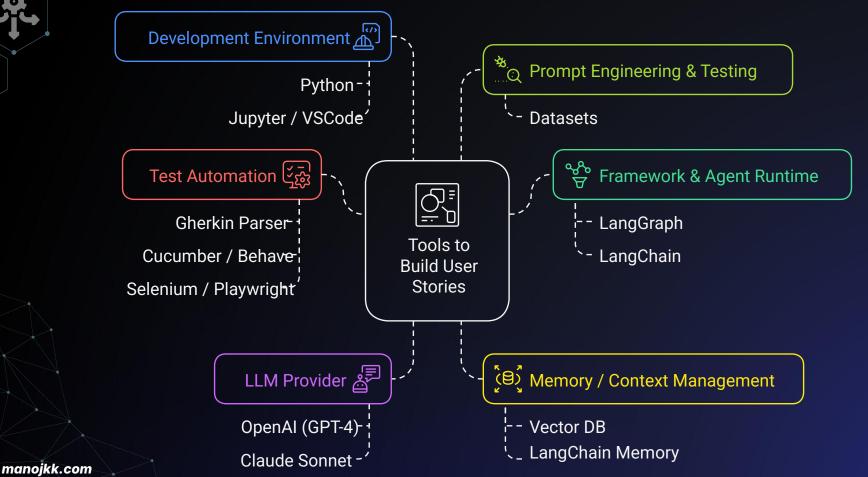


Review

Agent

The agent that reviews the generated content.

Agentic Al Toolkit





"It's Show Time!!"

Reference Architecture



Challenges...

Explainability

Understanding agent decisions is crucial.

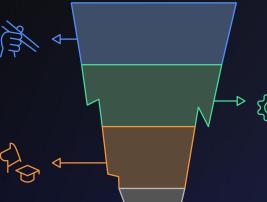


Control

Maintaining appropriate oversight is essential.

Skills

Upskilling teams for agent management is important.



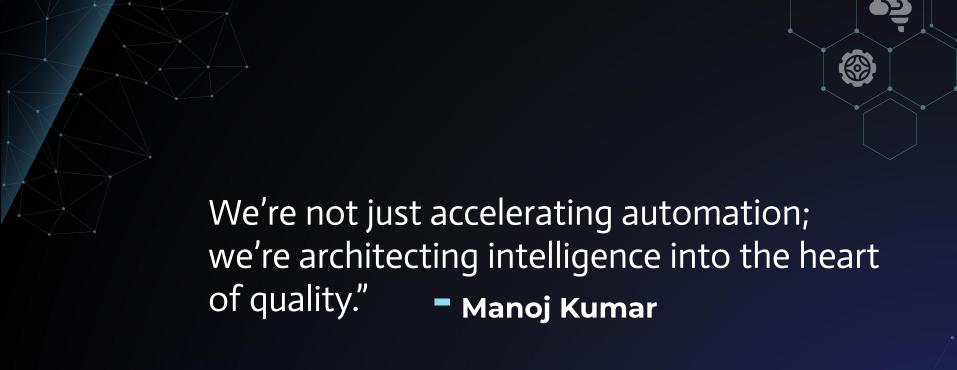
Integration

Working with existing systems is necessary.

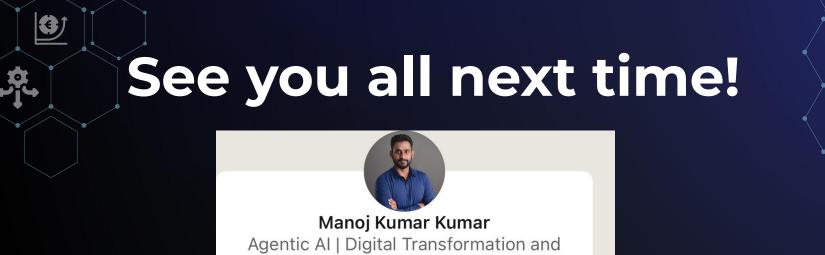


Ensuring responsible AI use is paramount.

manojkk.com













Wikehomakeef Argahtige Att 'Agentic'?



R&D Automation

Accessibility Auditing

DevOps



How Agentic Al works... Memory Mgmt...



Long Term: Vector databases store persistent agent knowledge and experiences

In-context Short-term Memory: Active working memory for current task execution

Context Sharing: Seamless information flow between tasks and execution periods

Made with ≽ Napkin

Evolution of Agentic Al

