Problem statement for launching an AI Capability Growth

Client expectation

- Deploy OOTB (Out of the Box) Al enablement capabilities internally to drive efficiency and effectiveness across all of organization.
- There is absolutely no excuse for not making AI assisted development part of your workflow. With all the productivity evidence and free assistance and training provided, I would not even consider engineers for a job without skills and experience using AI dev tools. Invest time in your career.

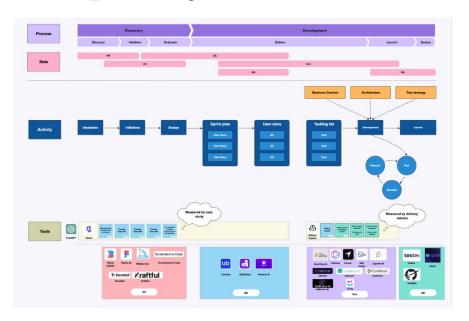
Industry Trend

We believe there is an opportunity for TWs delivery teams to leverage AI tooling to help us deliver more efficiently to assist our clients



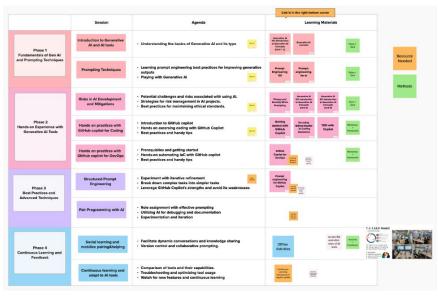
2024 Thomas tingetting value to market faster.

AI Capability Growth Framework



Building AI Capability Growth framework based on

Discovery-and-Delivery-Process to solve #1, #3, #4, #5, #7. It allows us to look at the integration of AI in each activity from an agile engineering perspective, and the potential tooling options also allow us to keep our eyes on development trends.



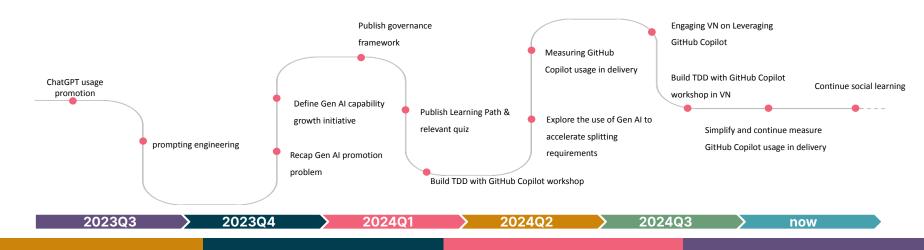
From a practical perspective, a learning path map is developed to enable individuals to learn the fundamentals, hands-on practice and trendy development directions end-to-end to address #2, #5, #6, #7

Al Capability Growth Roadmap

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		Session	Agenda	Link is in the right bottom corner Learning Materials
	Phase 1 Fundamentals of AI and Language Models	Introduction to Artificial Intelligence	Overview of AI, machine learning, and deep learning. Differences between narrow AI and general AI. Fundamentals of Language Models. Introduction to Natural Language Processing (NLP).	An executive's guide to Al & ML Splunk Al & ML Section 1) An executive's Al & ML Section 3 - (Section 1) Machine Learning NLP Section 3 - (Section 1)
		Use Cases of Al in Software Development	Exploration of Al in the software development lifecycle. Real-world examples of Al implementation in software projects.	LLM empowers Al-Assisted Software Software Improvement Delivery Guideline Guideline
	Phase 2 Gen Al Data and Model Understanding	Generative AI and Data Model	Basic principles and mechanisms Model outputs and informed decision-making. Problems of biases and inaccuracies.	Generative M. St. Interdection to deterement AL Concepts (Unit 1 - 3) Debrind (Section 1 - 3) ChatGPT Generative AL Learn about the next All frontier actually works
	Phase 3 Risks, Ethics, and Mitigations	Risks in Al Development and Mitigations	Potential challenges and risks associated with using Al. Strategies for risk management in Al projects. Best practices for maintaining ethical standards.	Generative M. Generative M. SecConf2022: 101: Immodestion M. 101: Immodestion Modigating privacy and risks of Al. Security When concepts Concepts Imaguage Prompting model
	Phase 4 Effective Prompting	Effective Prompting Techniques	Comparison of tools and their capabilities. Crafting prompts to get desired outputs. Tips for refining prompts for better results. Troubleshooting and optimizing tool usage.	Prompt Prompt Camp Camp Camp Colkit of Al toolkit of Al techniques charges and the complete prompt of Al bootcamp Camp Camp Camp Camp Camp Camp Camp C
	Techniques	Collaborative Development with Using Gen Al	Team collaboration strategies when using Al tools. Version control and collaborative prompting.	GitHub MS ANS Copilot Copilot Copilot Codex
	Phase 5 Hands-on Experience with Generative Al Tools	GitHub Copilot (Dev Only)	Introduction to GitHub copilot Best practices and handy tips	Getting Decoding started with GilHub Copilot -Al Coding Copilot -Al Soding Assistance
		Hands-on with GitHub Copilot (Dev Only)	Prerequisites and getting started Hands-on coding with GitHub copilot	GitHub Copilot for Coding GitHub Copilot for DevOps
	Phase 6 Advanced Techniques and Integration	Fine-Tuning Models for Specific Domains	Tailored models for domain-specific requirements. Hands-on experience with fine-tuning.	LLM fine- tuning on OpenAl Translation
		Cloud Integration	Strategies for seamless integration of AI with cloud service. Automation and optimization techniques.	MS Copilot for Azure
© 2024 Though		Future Trends of Gen Al Plus Software Development	Exploration of emerging technologies and their impact on software development. Prediction on the future trajectory of Gen AI in the technology field.	Genal: Possibilities vs. Costs Generative AI Solution Exploration: PE, RAG, FT Repid LLM App Protobyping A No Code Approach with Dify ai

Our Journey of AI Capability Growth Initiative



> 90%

Finish 7 sets of quizzes for Learning path stage 1~4

80.5%(Xian) 61.9% (VN)

devs completed the TDD with Github Copilot workshop

100%

- Everyone who participates in the workshop submit at least one insight, including ticket, commit, etc
- Keep collecting insights from new member by team iteratively collecting GitHub Copilot usage data

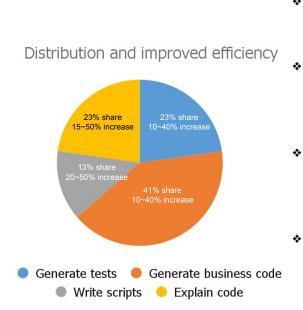
15.8%

Using GitHub Copilot reduced coding time for four major usage scenarios, reducing the average cycle time of each team by 15.8%

GitHub Copilot Applicability Analytics

Getting feedback from all team members that participated in the TDD with GitHub Copilot workshop. Picking up 1 of sprint data, these teams completed a total of 150 tickets in the recent iteration. Among them, 73 tickets did use GitHub Copilot (48.7%), 77 tickets did not use GitHub Copilot (51.3%).

The teams' usage scenarios can be roughly divided into four categories:



- Generate test: including generating test code and test data. Because the code structure is simple, the accuracy is higher.
- by the business code: the accuracy is limited by the business context, and the probability of the generated code being adjusted is relatively high, about 30% ~ 80%.
- Write scripts: Since the generated scripts are less relevant to the business, the prompting needs to contain less information and the generated code is more accurate.
- Explain code: Teams use code explanation as a more efficient way to collect business code and information, and some teams even use it to share business context in a session, especially for the legacy code.

Scenarios where GitHub Copilot is Not used:

- Local environment setup
- Manual tests
- Knowledge sharing
- Straightforward tasks
- Tools decommission or upgrade
- Designing and Solutioning
- **♦** Alerting & monitoring
- Supporting requests
- **♦** Vulnerability fixing
- Troubleshooting cross the system

GitHub Copilot Increases Speed and Productivity Our Thoughtworks Teams

Scenarios	Development time	Generate tests		Generate business code		Write scripts		Explain code		Applicable weighted	General
		Coding time saved	Cycle time saved	Coding time saved	Cycle time saved	Coding time saved	Cycle time saved	Coding time saved	Cycle time saved	cycle time	weighted cycle time saved 48.7%
Very optimistic	76.5% [X]	30.0% [Y i]	23.0%	50.0%	38.3%	50.0%	38.3%	40.0%	30.6%	33.0%	16.1%
Middle	55.4%	28.0%	15.5%	29.4%	16.3%	38.7%	21.4%	22.0%	12.2%	15.8%	7.7%
Pessimistic	48.0%	15.0%	7.2%	10.0%	4.8%	33.3%	16.0%	10.0%	4.8%	6.8%	3.3%

$$S = \sum_{i=1}^n X imes Y_i imes Z_i \quad n \in \{1,2,3,4\}.$$

48.7%: According to the most recent iteration, the team has completed about 150 tickets in total. Of these, 73 tickets did use GitHub Copilot (48.7%).

X = Development Time (the percentage of time the team spends on implementation and testing during a sprint. In the table above, this value represents the average for the Xi'an team in each segment over the past two months)

- Y = Coding Time saved (the percentage of coding time saved based on the analysis of each ticket collected by the team capability champion)
- **Z** = Usage Scenario Proportion (a statistical value representing the proportion of tickets for this type of usage scenario in the sprint)
- S = Applicable Cycle Time Saved (the average improvement rate of overall cycle time across teams after weighting each usage scenario (currently 4 usage scenarios))