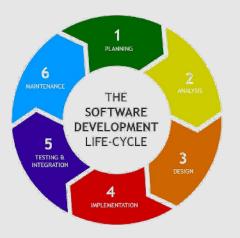


Incorporating AI into your SDLC

Leveraging AI tooling across the phases of your software development lifecycle

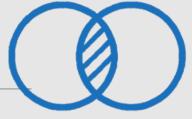












Presented by **Brent Laster**

Tech Skills Transformations LLC

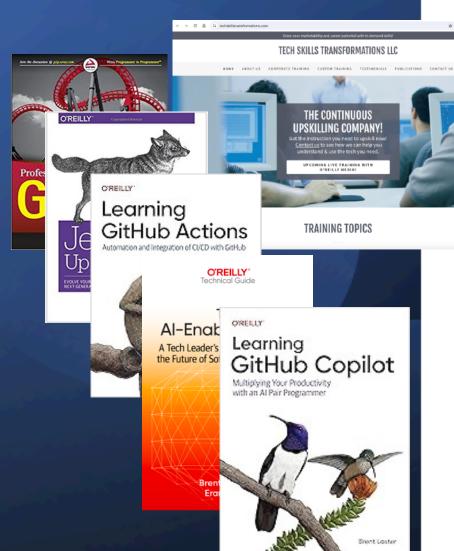
Al in Production

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About me



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- https://getskillsnow.com
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- Long career in corporate as dev, manager, and director in DevOps and other areas
- Author

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- O'Reilly "reports"
- **Books**
 - Professional Git
 - Jenkins 2 Up and Running
 - Learning GitHub Actions
 - AI-Enabled SDLC
 - Learning GitHub Copilot
- Speaker
- Social media
- LinkedIn: brentlaster
- X: @BrentCLaster
- Bluesky: brentclaster.bsky.social
- GitHub: brentlaster















Visual Studio Code



ChatGPT

LISTED BELOW ARE A FEW OF THE TECHNOLOGIES FOR WHICH WE OFFER HANDS-ON TRAINING. THESE CAN BE **CUSTOMIZED FOR ANY** SIZE TEAM FROM 1-100 AND FOR ANY LEVEL FROM BEGINNER TO ADVANCED.

- ArgoCD

- Git

- GitHub Foundations

- Grafana
- Helm
- Kubernetes
- Kustomize
- LLMs
- Prometheus
- Tekton
- VS Code



- Containers
- Docker
- Gerrit
- GitHub Actions
- GitHub Codespaces
- GitHub Copilot
- GitHub Security
- GitLab
- GitOps • Gradle

- Jenkins



Jenkins

kubernetes

Prometheus

docker

argo

Grafana

GitHub Actions















Background

Rise of AI Coding Assistants (Software Dev Asst's = SDA)

 LLMs and AI apps have exploded since OpenAI added chat to its models in November 2022



 Code creation and related are excellent fits for Gen AI due to structure, syntax, and wealth of examples



At core, provide inline code suggestions as well as NL chat interfaces

- But also provide additional powerful functions
 - Repository indexing (RAG)
 - Summarizing (PRs, Issues, etc.)
 - Agent workflows (From big ideas to complete implementation)

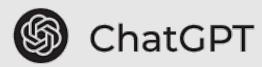




- Users can leverage them in three modes:
 - Acceleration
 - **Exploration**
 - Directive







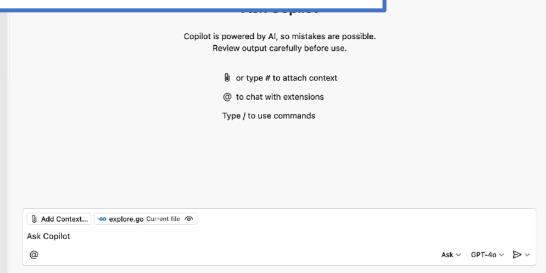


7 C + 5 ··· ×



- Accelerators/Acceleration code completion working in the flow and AI provides completion suggestions (usually in the editor)
- Explorers/Exploration using the chat interface to ask how to code something? What is the API?
- Directors/Di Specific instructions to create something via
- the chat interface or in the editor (via comments)
- All of these can lead to increased developer productivity

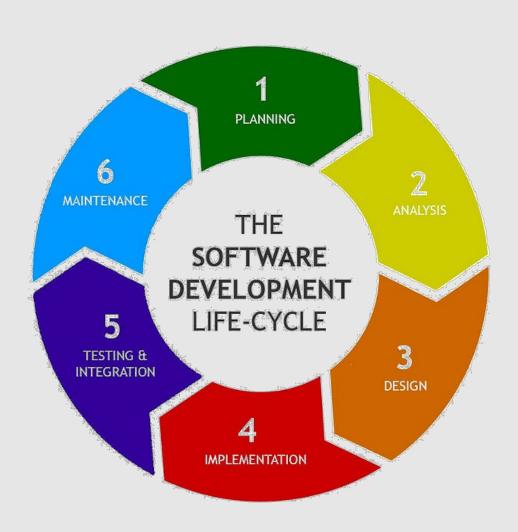
source: "https://www.mckinsey.com/capabilities/mckinsey-digital/our-insights/unleashing-developer-productivity-with-generative-ai





The Software Development Lifecycle (SDLC)

- Structured process used to create high-quality software efficiently and consistently
- Outlines and defines the steps to plan, develop, test, and deploy
- Typical phases include
 - Planning
 - Analysis
 - Design
 - Implementation
 - Testing and integration
 - Maintenance
- Defined phases may vary across organizations/projects



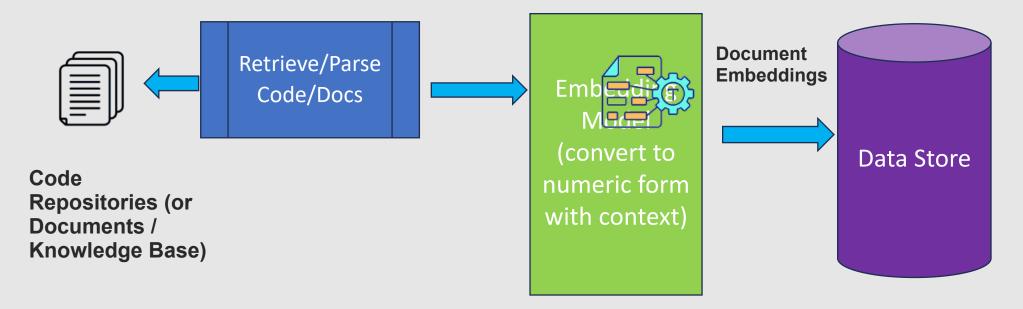


Planning



Background: RAG

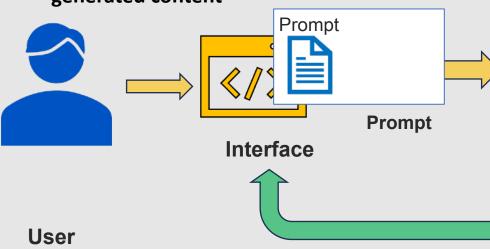
Goal: Search your local data (code, docs, etc.) to give additional context for LLM to work with and give you more targeted, useful responses (considers your data along with its training)



- Your data is parsed and stored with information about other data it's related to in a data store
 - parsing = tokens, storing as numeric data = embeddings, information about other data = vectors (set of numbers that represent how much a term is related to others in the data store)

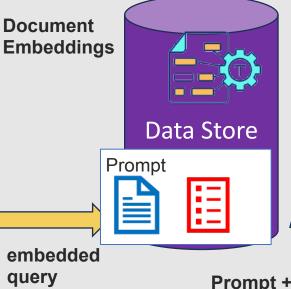
Background: RAG

- For queries/prompts, application gathers results (most relevant ones) from the vector database with your data
- Adds results to your regular LLM query/prompt
- Asks the LLM to answer based on the augmented/enriched query/prompt
- NOTE: Items returned via RAG search are existing items from the data store, not generated content



User Query and Response Generation

Embedding Model



Original prompt + matching "docs" (aka "enhanced context")

Prompt + enhanced context

LLM Response

LLM

.____

response (generative)



Planning with AI (approaches)

- Index your code (targeted RAG) usually automatic now
- Make code repositories accessible for AI (via RAG/indexes)
 - Easily find and identify reusable code
- Ask questions of your code to simplify estimates, avoid rework, ground knowledge
 - How did we implement function X?
 - Does our code use dependency Y that has this security vulnerability? If so, where?
 - We want to migrate from one platform/cloud/language / framework to another where do we need to focus our efforts?
 - The open-source library we use has changed licensing terms how are we using that?
 - Do we have code that does Z?



Planning: Prototyping

- For brainstorming/considering requirements
- Can quickly draft pseudo code in comments
- Al can generate "draft" code
- Doesn't have to be perfect, just enough for estimation/planning

```
# Given a list of app features and their dependencies as (feature, depends_on) pairs
# Determine a valid order in which the user can unlock and access features
# Raise an error if a cycle exists (invalid dependency loop)
# Example input:
      features = [
          ("upload", "signup"),
          ("share", "upload"),
          ("analytics", "share"),
          ("signup", None)
  Expected output:
      ["signup", "upload", "share", "analytics"]
# Define a function to accept the feature dependency list
# Build a directed graph using adjacency lists
# Use topological sort to determine feature unlock order
# Validate and raise an error if dependencies form a cycle
# Return an ordered list of unlockable features
```

```
from collections import defaultdict, deque
def determine feature unlock order(dependencies):
    # Build graph and in-degree count
    graph = defaultdict(list)
    in degree = defaultdict(int)
    features = set()
    for feature, depends_on in dependencies:
        features.add(feature)
        if depends on:
            graph[depends_on].append(feature)
            in degree[feature] += 1
            features.add(depends on)
        else:
            in degree[feature] += 0 # Ensure entry in in-degree map
    # Start with features that have no dependencies
    queue = deque([f for f in features if in_degree[f] == 0])
    unlock order = []
    while aueue:
        current = queue.popleft()
        unlock order.append(current)
        for dependent in graph[current]:
            in degree[dependent] -= 1
            if in_degree[dependent] == 0:
                queue.append(dependent)
    if len(unlock_order) != len(features):
        raise ValueError("Circular feature dependency detected")
    return unlock_order
# Example usage
if __name__ == "__main__":
    features = [
        ("upload", "signup"),
        ("share", "upload"),
        ("analytics", "share"),
        ("signup", None)
    order = determine_feature_unlock_order(features)
    print("Feature unlock order:", order)
```



Development



Types of Al Assistance for Coding

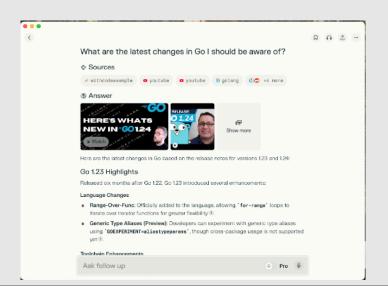
 Extensions that add AI code completions and chat





- Al Coding IDEs
 - Based on VS Code
 - Al-first design
 - Tend to focus more fully on project-wide insights and application building
 - Think "collaborate like Copilot and handle complex tasks like an agent"
- Research-focused Al assistants
 - Offer answers based in current content
 - Synthesize info from multiple sources (cited) into concise, natural language answers
 - Uses LLM on back end
 - Great at real-time info retrieval
 - Can also do most things that ChatGPT can do (but with up-to-date info and sources)









How coding assistants can help and what to remember

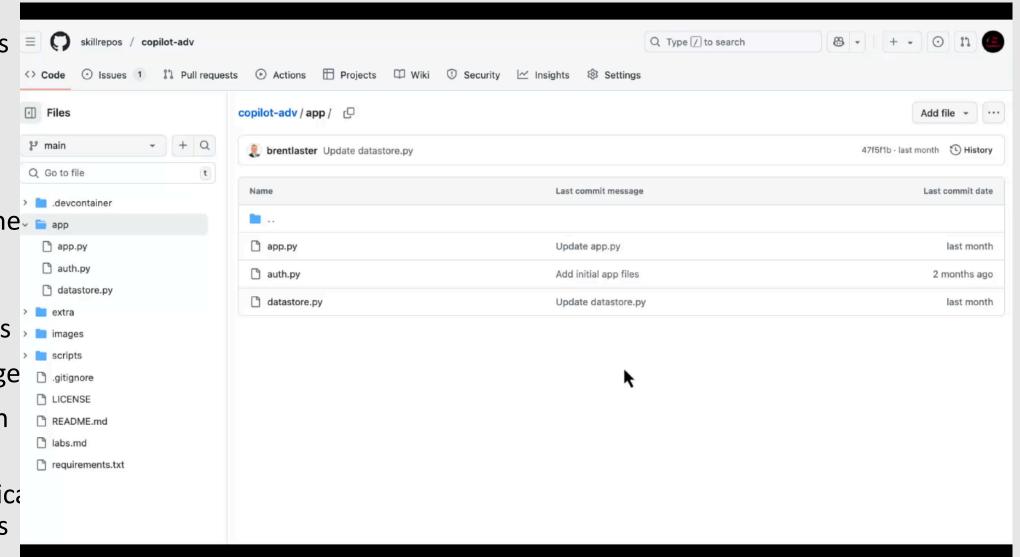
- Code suggestions for nearly any current language
- Comment driven code creation
- Automatic test generation
- SQL generation (for queries, tables, indices, stored procs)
- Regular expression generation
- Pattern/mappings generation
- Document code
- Explain code
- Fixing code
- Reviewing code
- Translating code
- Generate specs/configs
- What is the API to do...

- Reduce cognitive load
- > Let programmers focus on the parts that need their expertise
- ➤ Al needs good context
- ➤ AI needs good guidance or queries (prompts)
- ➤AI can hallucinate
- ➤ AI can be out-of-date
- ➤ AI doesn't do real-time validation of generated data
- ➤AI doesn't care about tech debt
- ➤ BOTTOM LINE: Al-generated code intended for production must ALWAYS be subject to same oversight and review



Leveraging Agents

- Most Al assistants have agentic functions built in
- Agent = Al w/ agency
 - ability to gathe input
 - act
 - make changes
- End-to-end change
- Autonomous with oversight
- Goes beyond typical
 Al assist functions





Resolving Issues



Resolving issues with Al

- Shortcut fix commands available
- Prompts to help find and fix issues (or potential issues)
 - What are some places in this repository where logic/performance/security errors can occur?
 - What are the most recent significant changes to the codebase?
 - What are the code changes that affected this code in branch x from date y to date **z**?
 - How can I make this code be more efficient with system resources?
 - Where else might this problem occur?
 - What other parts of the codebase use this routine?

```
Tabnine Protected
 opts := zap.Options{
    Development: true,
 opts.BindFlags(flag.CommandLine)
 flag.Parse()
 In the provided code, the errors were caused by incorrect
variable declaration and incorrect usage of the zap.Options
and flag packages. The opts variable should be declared
```

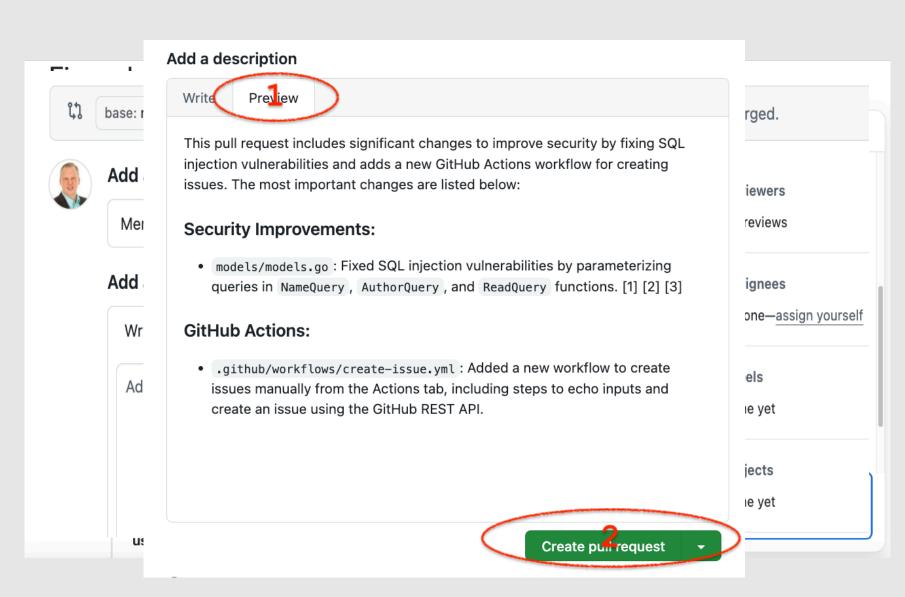
as a zap.Options type, and the flag.CommandLine should

```
prime.py >  is_prime
      def is prime(n):
          if n \le 1 or (n > 3 and (n % 2 == 0 \text{ or } n % 3 == 0)):
 3
              return False
       /fix "x" is not defined
            > Used 1 reference
           The issue is that the variable of x is undefined; it should be
           replaced with on, which is the parameter of the function.
       Ask Copilot
                                                     Accept
               Close ♡ ∨
 5
          if n <= 3:
 6
              return True
```



Working with tracking mechanisms (Copilot examples)

- Given an issue, can prompt Copilot for next steps
- Prompt to provide summary of issue
 - Includes overview of issue based on what Copilot can gather from the system using tools
- Generate pull request summaries





Al review – Having Copilot review code in VS Code



- Al-powered feedback on code as you write it
 - Review selection:
 - » select code in VS Code and ask for initial review
 - » Available now for all Copilot subscribers

- Review changes: ask Copilot for deeper review of all changes before pushing from "Source Control" tab
 - » Can also be done in pull requests on GitHub.com
 - » Currently must be on waitlist



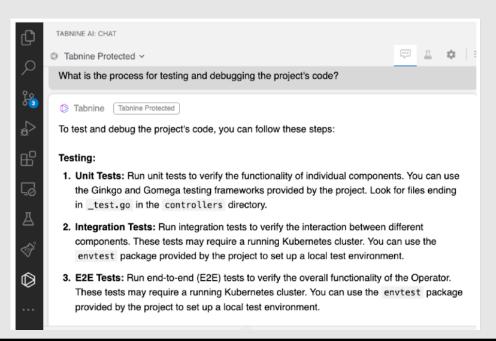


Testing

Testing with AI

- Create tests for ...
- What is the process for testing and debugging the project's code?
- Generate unit tests / integration tests for

. . .



```
TABNINE AI: CHAT
Tabnine Protected >
 Create tests for the `is_prime` function using Python.
             Tabnine Protected
 Tabnine
  def test_is_prime():
      assert is_prime(2) == True
      assert is_prime(3) == True
      assert is_prime(4) == False
      assert is_prime(5) == True
      assert is_prime(6) == False
```

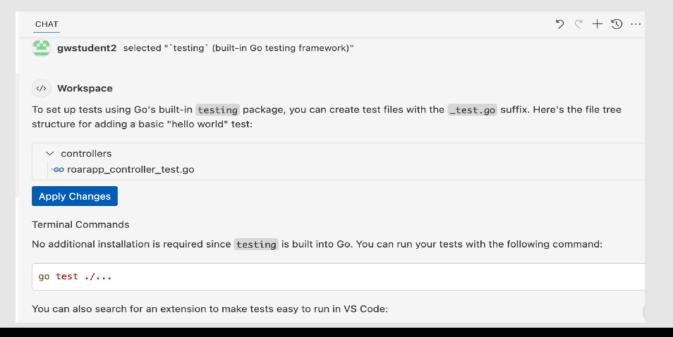
```
prime.py > 分 is_prime
                                                               GPT 4o ∨ ⊳
    /tests
                                     Generate unit tests for the selected code
          /tests
      🛼 if n == 1:
      ····return False
       for i in range(2, int(n ** 0.5) + 1):
       ····if n % i == 0:
       ····return False
      · · · return True
```

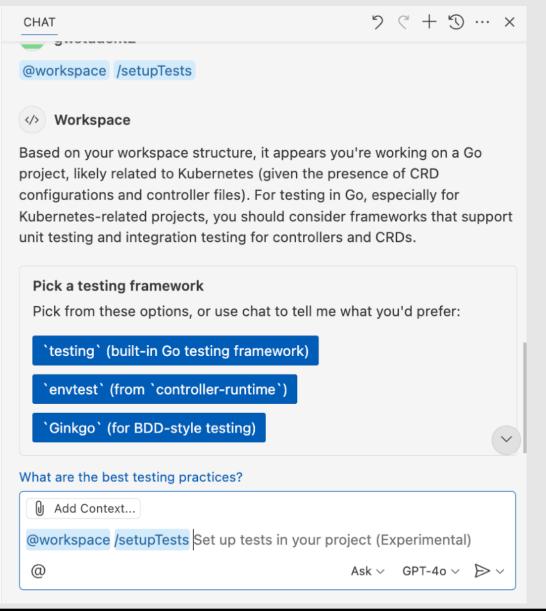




Al Testing – from the start

- Can help set up a testing framework for your code
- Recommend appropriate frameworks
- Provide step-by-step setup and configuration instructions
- Suggest relevant testing extensions for your dev env
- Can have custom test generation instructions





+ 50 ...



Testing – beyond the basics

- What other edge cases should I be testing?
- How do I test X for performance?
- How do I test Y for security?
- Test Driven
 Development
 - Create dependency file
 - Create test class (fail)
 - Create implementation class (succeed)

AT

- 1. Zero and One: Both are not prime numbers by definition, so it's important to test these explicitly.
 - Negative Numbers: While the definition of prime numbers typically does not include negatives, ensuring your function gracefully handles negative inputs is good practice.
 - 3. Large Prime Numbers: Testing with large prime numbers can help ensure your algorithm performs well under stress and correctly identifies primes in all cases.
 - 4. Large Non-Prime Numbers: Similarly, large non-prime numbers, especially those that are products of two large primes, can be challenging for some algorithms.
 - 5. **Very Small Primes**: The first few primes (2, 3, 5, 7) are often corner cases in prime number algorithms, especially if any optimizations or shortcuts are taken.
 - 6. **Sequential Primes**: Testing sequences of prime numbers (like 11, 13, 17, 19) can help identify issues with algorithms that might have patterns or periodic errors.
 - 7. **Primes That Are Products of Small Primes**: Numbers like 49 (7x7), 121 (11x11), or 169 (13x13) can test if the algorithm correctly identifies squares of primes as non-prime.
 - 8. **Floating Point Numbers**: While prime numbers are inherently integers, ensuring your function correctly handles or rejects floating-point numbers is a good practice.
 - 9. **Data Types Other Than Integers**: Passing strings, lists, or other non-integer types can help ensure your function is robust against type errors.

Here's how you might extend the test suite to cover some of these cases:

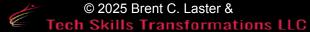
```
def test_is_prime_zero():
    assert is_prime(0) == False

def test_is_prime_one():
    assert is_prime(1) == False
```

What are some common optimization techniques for prime number checking algorithms?

are there any other edge caes that should be tested?







Creating Tests Via Prompts

• Prompt for Unit Tests:

"Develop a comprehensive suite of unit tests for the <method/function> in <language/framework>. Write multiple test methods that cover a wide range of scenarios, including edge cases, exception handling, and data validation."

Prompt for Integration Tests:

"Write integration tests for the <method/function> in the <class>. Use mocks to simulate the <external service> and verify that it works correctly when <condition>."

Source: https://docs.github.com/en/copilot/using-github-copilot/example-use-cases/writing-tests-with-github-copilot#writing-unit-tests-with-copilot-chat



Refactoring/Updating



Refactoring with AI

Consider a code example

```
    Can refactor via prompt
```

Resulting code

```
def is_prime(n):
    return n > 1 and all(n % i for i in range(2, int(n**0.5) + 1))
```

The is_prime function should have more meaningful variable names (max of 3 characters) and comments for better readability.

```
# Check if the input is a positive integer greater than 1
    if not isinstance(num, int) or num <= 1:
        raise ValueError("Input must be a positive integer greater
than 1")

# Check if the number is greater than 1 and not divisible by
any number up to its square root
    if num > 1 and all(num % div for div in range(2, int(num**0.5))
+ 1)):
        return True
    else:
        return False
```

3 Q



Refactoring for better testability

- Can also ask Copilot to refactor code to be more suitable for testing
- For example: "Refactor the code in #file:<filename> to make it more easily testable"

```
def is_prime(n):
    if not isinstance(n, int):
        raise ValueError("Input must be an integer")
    if n < 0:
        raise ValueError("Input must be a non-negative integer")

if n < 2:
    return False
    for i in range(2, n):
        if n % i == 0:
            return False
    return True</pre>
```



> Used 1 reference

To make the code more testable, we can refactor it by separating the validation logic from the core functionality. This way, we can test the validation and the prime-checking logic independently. Here is the refactored code:

```
def validate_input(n):
    if not isinstance(n, int):
        raise TypeError("Input must be an integer")
    if n <= 1:
        raise ValueError("Input must be greater than 1")

def is_prime(n):
    validate_input(n)

    if n == 1:
        return False
    for i in range(2, int(n ** 0.5) + 1):
        if n % i == 0:
            return False
    return True</pre>
```

Now, we can write separate unit tests for validate_input and is_prime functions. This separation of concerns makes the code more modular and easier to test.



Exploratory Refactoring and Updating with Al

- Can probe with questions to decide what needs to be updated?
 - Where do we use API v1alpha1?
 - What are the changes between v1alpha1 and v1 API versions?
- Can prompt to do updates such as "Change v1alpha1 API to v1"
- Al-recommended change areas
 - import paths
 - references to API versions
 - scaffold comments
 - Controller code

- Tabnine Protected
 - 3. Update the scaffold comments:
 - Replace //+kubebuilder:scaffold:imports with //+kubebuilder:scaffold:imports:v1.
 - Replace //+kubebuilder:scaffold:scheme:v1.
 - 4. Update the controller code:
 - If you have any custom logic or code that references the v1alpha1 version of the API, update it to use the v1 version.

Here's an example of how the updated code might look like:

```
import (
    "flag"
    "os"

// Import all Kubernetes client auth plugins (e.g. Azure, GCP, OIDC, etc.)
// to ensure that exec-entrypoint and run can make use of them.
"k8s io/client-go/plugin/pkg/client/auth"
```



Augmenting Code with Al

- Augmenting = adding additional functionaly to existing code
- Prompt "The is_prime function should log whether the in put is prime or not"
- Al suggested additional code that
 - pulls in a logging framework, initializes a logger instance
 - adds logging calls for each possible output

logging module in Python. Here's an example of how you can do it:

```
import logging
logging.basicConfig(level=logging.INFO)
logger = logging.getLogger(__name__)
def is_prime(n):
    if not isinstance(n, int) or n <= 1:
        raise ValueError("Input must be a positive integer greater than 1")
    if n > 1 and all(n % i for i in range(2, int<math>(n**0.5) + 1)):
        logger.info(f"{n} is a prime number")
        return True
    else:
        logger.info(f"{n} is not a prime number")
        return False
告 Insert □ Copy 元 New file
```

In this updated version, the logging.basicConfig(level=logging.INF0) line sets up the





Changing Implementations with Al

- Can select code and prompt for alternative implementations
- Consider example code
- Prompt "Explain this function in detail and suggest alternative implementations"
- Also porting
- Prompt "convert to Go"

```
package main
import "fmt"
func fibonacci(n int) []int {
    sequence := []int{0, 1}
    for sequence[len(sequence)-1] + sequence[len(sequence)-2] <= n {
        sequence = append(sequence, sequence[len(sequence)-1] +
sequence[len(sequence)-2])
    return sequence
func main() {
    n := 100
    fmt.Println(fibonacci(n))
```

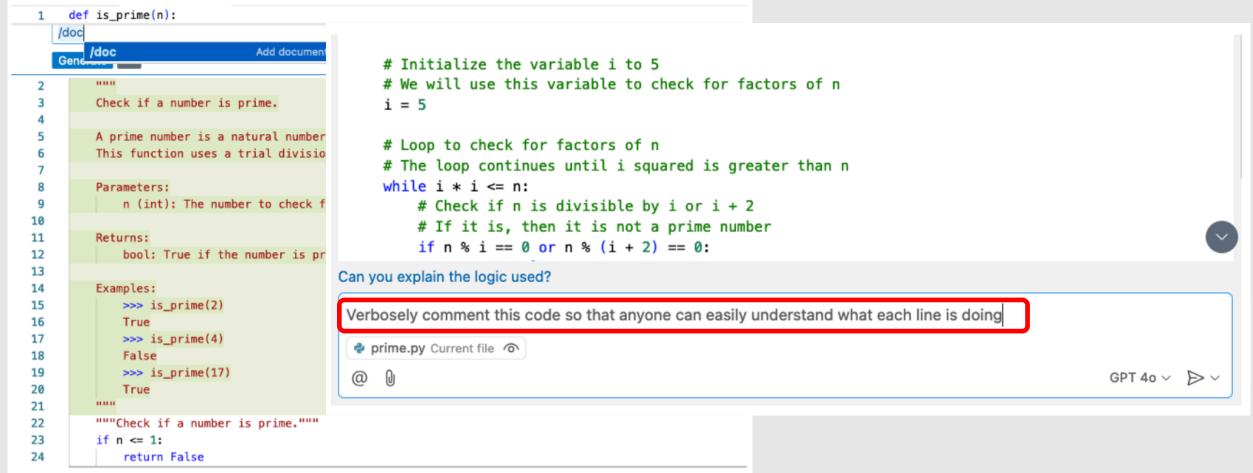


Documentation



Using AI for general documentation via comments

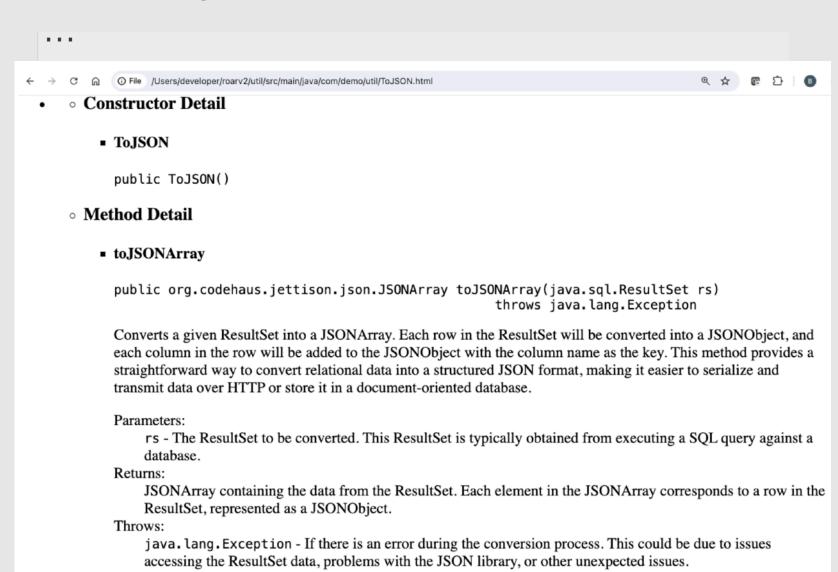
- Basic methods may just do "header" comments
- By being more specific in prompt, can get more extensive comments





Generating Framework-compatible documentation

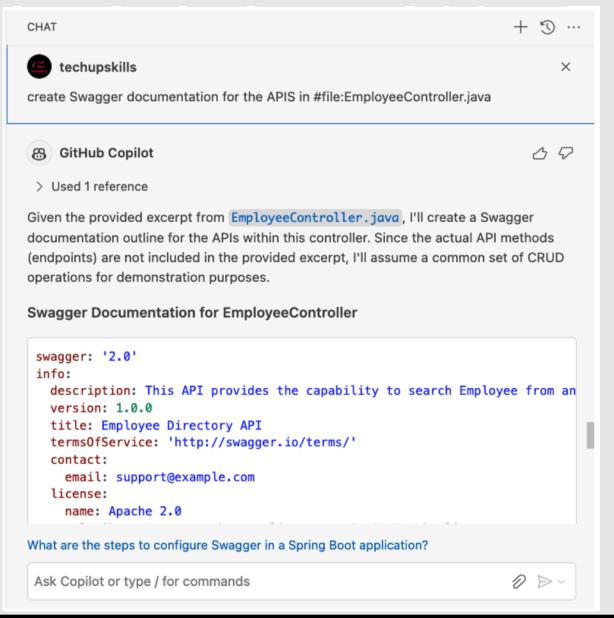
- If the SDA recognizes that the code is in a language that has a standard documentation tool associated with it, it will default to creating comments ready for that tool
- Given example code...
- Have Copilot document it
- Can then generate framework doc from it





Generating Documentation for APIs

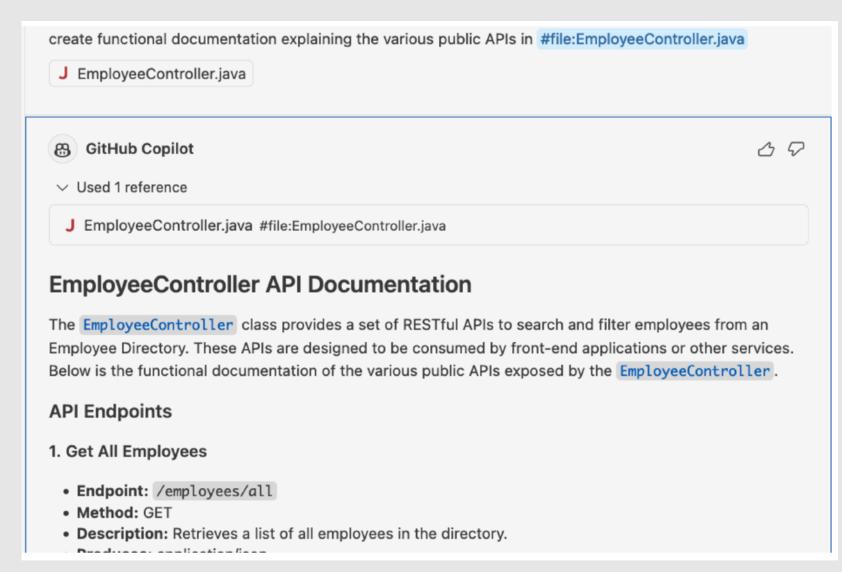
 If your codebase is setup for an API documentation framework like <u>Swagger</u>, then you can have the SDA generate the corresponding Swagger documentation for your APIs





Generating Functional Documentation

- SDAs are also capable of creating functional documentation that is targeted towards external users
- May also produce a set of related info including Identifying Public APIs, Method Signatures, Endpoint Mapping, etc.







Extracting Summary Documentation

- Copilot can also be used to extract a high-level summary of the key documentation from certain types of projects
- This can be extremely useful for quickly understanding the significant details of the code base for a project

The key documentation extracted from the provided code and workspace structure highlights the setup and configuration of a Spring Boot application with Swagger for API documentation. Here's a summary:

Project Structure and Configuration

- Project Base: Spring Boot application designed for demonstrating Swagger integration.
- Group ID: com.techupskills.springboot.swagger
- Artifact ID: springboot-swagger
- Version: 0.0.1-SNAPSHOT
- Java Version: 17
- Swagger Version: 3.0.0
- Packaging: JAR
- Parent: Spring Boot Starter Parent with version 3.1.0
- Source Encoding: UTF-8

Dependencies

- Spring Boot Starter Web: For building web, including RESTful, applications using Spring MVC.
- Spring Boot Starter Test: For testing Spring Boot applications with libraries including JUnit, Hamcrest, and Mockito.
- Spring Boot Starter Data JPA: For using Spring Data JPA with Hibernate.

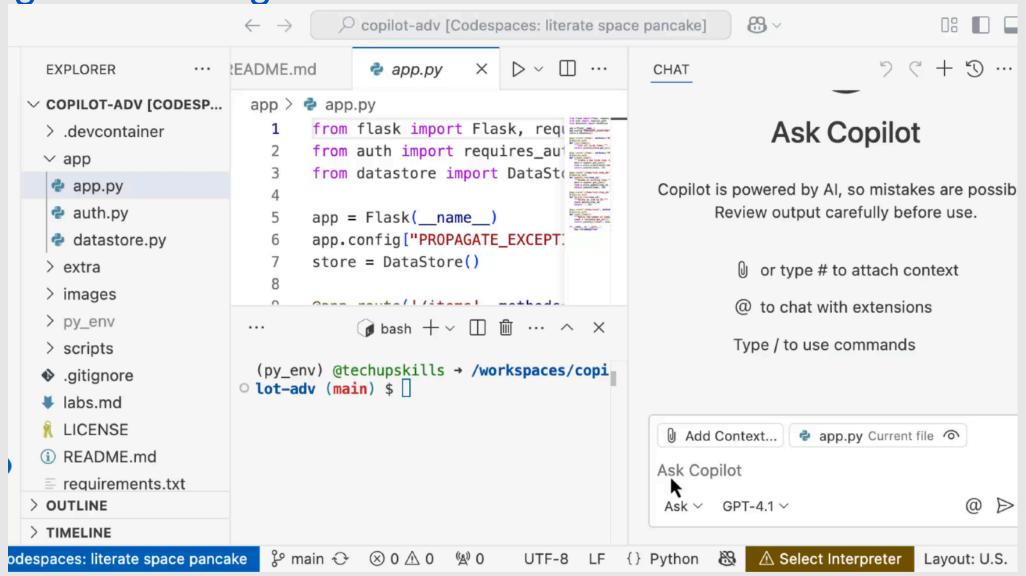




Simplifying Maintenance w/ Explaining/Onboarding

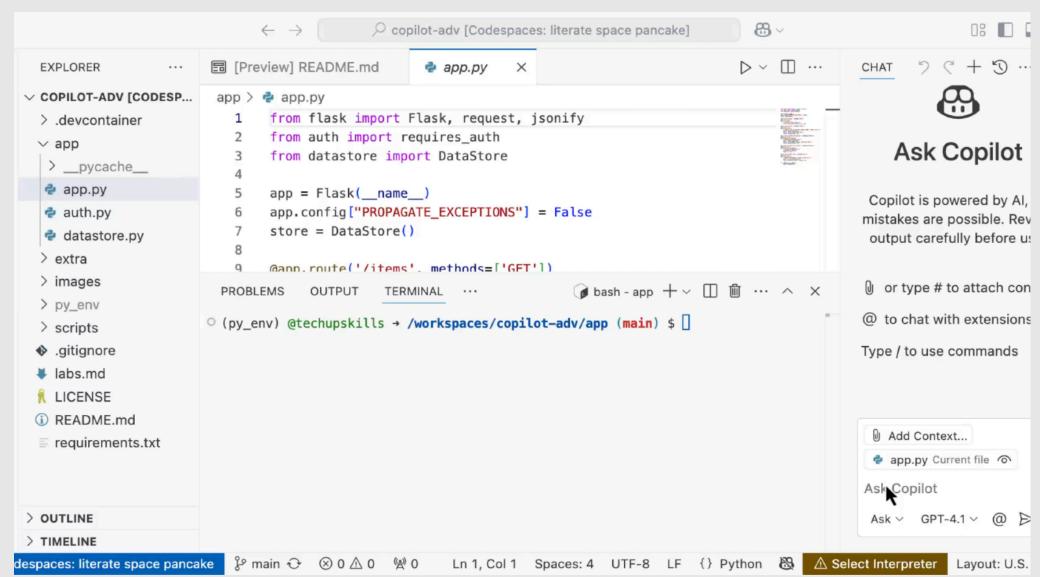
Simplifying Onboarding

- Onboarding is a key aspect of helping devs get going with code
- Simple to have Al generate an onboarding guide





- Key for taking over/assisting with code bases
- Useful for learning new frameworks/ languages
- Using built-in "explain" functionality
- Can also have it explain how to see/run functionality



Contact: training@getskillsnow.com

techskillstransformations.com getskillsnow.com

