How artificial intelligence can help and hinder the programmer.
A case study.

Marek Bryling



Disclaimer

Most of the slides in this presentation were generated by ChatGPT and translated into English by DeepL.



About the author

Marek Bryling

Practice Lead at STX Next

- Team support
- Introducing new practices and technologies
- Improving knowledge sharing
- Overseeing R&D projects
- Company-wide engagement
- 30 Years of IT industry experience



"

- "Code churn," or the percentage of lines thrown out less than two weeks after being authored, is on the rise and expected to double in 2024. The study notes that more churn means higher risk of mistakes being deployed into production.
 - The percentage of "copy/pasted code" is increasing faster than "updated," "deleted," or "moved" code. "In this regard, the composition of AI-generated code is similar to a short-term developer that doesn't thoughtfully integrate their work into the broader project,"

GitClear founder Bill Harding



Introduction

Definition of AI in the context of a programmer's work

https://chat.openai.com/, https://copilot.microsoft.com/, https://codium.ai/, https://codium.ai/, https://coderabbit.ai/









- Automate routine tasks such as testing code, generating reports,
 and optimizing processes
- Increase efficiency through data analysis and trend forecasting
- Facilitate decision-making with machine learning algorithms



- Automate routine tasks such as testing code, generating reports,
 and optimizing processes
 - Code hinting
 - Code prompting
 - Code generation
 - Code optimizations
 - Finding bugs in code
 - Using new and unfamiliar technologies
- Increase efficiency through data analysis and trend forecasting
- Facilitate decision-making with machine learning algorithms





- Automate routine tasks such as testing code, generating reports,
 and optimizing processes
- Increase efficiency through data analysis and trend forecasting
 - Code profiling
 - Optimizations
 - Finding vulnerabilities
- Facilitate decision-making with machine learning algorithms



- Automate routine tasks such as testing code, generating reports,
 and optimizing processes
- Increase efficiency through data analysis and trend forecasting
- Facilitate decision making with machine learning algorithms
 - Automated code review
 - A new solution to the problem
 - Suggestions for new libraries
 - o I don't know how to do it, AI will tell me



How AI can spoil into a programmer's job

- The risk of errors in data interpretation and algorithm performance
- Potential for AI-based systems to replace the work of programmers
- Potential limitation of programmers' creativity and inventiveness through dependence on smart tools



How AI can spoil into a programmer's job

- It is estimated that for about 30-50% 30-33%* of junior-level developers,

 All can be a problem when working with more advanced applications.
- The customer doesn't want AI, but the developer can't do without AI
- The generated code is of poor quality
 - Code review hell
 - Prompting hell
 - Security issues
 - Unreadable code
 - o more ...





Research

Coding on Copilot: 2023 Data Suggests Downward Pressure on Code Quality

2023 Data Shows Downward Pressure on Code Quality

150m lines of analyzed code + projections for 2024

... GitHub's research finding on this point says "developers are 75% more fulfilled when using Copilot." ...

Al is like a "brand new credit card here that is going to allow us to accumulate technical debt in ways we were never able to do before," Armando Solar-Lezama, a professor at MIT

Research

The Impact of AI Tool on Engineering at ANZ Bank An Empirical Study on GitHub

Copilot within Corporate Environment, Blog post with a summary

The findings were notable: the group using Copilot completed their tasks 42.36% faster than the control group. Specifically, the control group took an average of 30.98 minutes per task, while the Copilot group averaged 17.86 minutes.

Case study

Create a Case class that can be initialized from data like the example(s) below.

Add functionality whereby the elapsed open time (duration) for any Case

may be retrieved (in seconds).

Case study (ChatGPT 3.5)

```
from datetime import datetime, timezone
    class Case:
         def __init__(self, id, author, created_date, closed_date):
             self.id = id
             self.author = author
             self.created_date = datetime.fromisoformat(created_date)
             self.closed_date = None if closed_date is None else_
                 datetime.fromisoformat(closed_date)
         def duration(self):
12
             if self.closed_date is not None:
                 return (self.closed_date - self.created_date).total_seconds()
             else:
                 return (datetime.now() - self.created_date).total_seconds()
```

Case study - test (ChatGPT 3.5)

```
open_case_dict = {
         "id": 101,
         "author": "ABCD",
11
         "created_date": "2020-01-01T00:00:00+06:00",
         "closed_date": None,
12
13
15
     closed_case = Case(**closed_case_dict)
     open_case = Case(**open_case_dict)
     print(closed_case.duration()) # Output: 12600.0 (3.5 hours in seconds)
     print(open_case.duration())
                                    # Output: 946080.0 (11 days in seconds)
20
```

Case study (ChatGPT 3.5-turbo)

```
from datetime import datetime, timezone
     class Case:
         def __init__(self, id, author, created_date, closed_date):
             self.id = id
             self.author = author
             self.created_date = datetime.fromisoformat(created_date)
             self.closed_date = None if closed_date is None else_
                 datetime.fromisoformat(closed_date)
         def duration(self):
12
             if self.closed_date is not None:
                 return (self.closed_date - self.created_date).total_seconds()
             else:
                 return (datetime.now(timezone.utc) - self.created_date).total_seconds()
```

Case study (ChatGPT 3.5-turbo)

```
24
     open_case_dict = {
         "id": 101,
         "author": "ABCD",
         "created_date": "2020-01-01T00:00:00+00:00",
         "closed_date": None,
     closed_case = Case(**closed_case_dict)
     open_case = Case(**open_case_dict)
     print(closed_case.duration()) # Output: 43200.0 (12 hours in seconds)
     print(open_case.duration()) # Output: 7571167.332743 (elapsed time since creation
37
```

Case study - test (ChatGPT 3.5-turbo)

```
open_case_dict = {
         "id": 101,
         "author": "ABCD",
11
         "created_date": "2020-01-01T00:00:00+06:00",
         "closed_date": None,
12
13
15
     closed_case = Case(**closed_case_dict)
     open_case = Case(**open_case_dict)
     print(closed_case.duration()) # Output: 12600.0 (3.5 hours in seconds)
     print(open_case.duration())
                                    # Output: 946080.0 (11 days in seconds)
20
```

Case study (ChatGPT 4)

```
from datetime import datetime, timezone
import pytz
class Case:
    def __init__(self, case_dict):
        self.id = case dict["id"]
        self.author = case_dict["author"]
        self.created date = datetime.fromisoformat(case dict["created date"])
        self.closed_date = datetime.fromisoformat(case_dict["closed_date"]) \
            if case_dict["closed_date"] else None
    def get_elapsed_open_time(self):
        """Returns the elapsed open time in seconds."""
        end_time = self.closed_date if self.closed_date else datetime.now(timezone.utc)
        return (end_time - self.created_date).total_seconds()
closed case = Case(closed case dict)
open_case = Case(open_case_dict)
# Get the duration for the closed case
print(f"Closed case duration: {closed_case.get_elapsed_open_time()} seconds")
print(f"Open case duration: {open_case.get_elapsed_open_time()}} seconds")
```



The Future

- DeepNext <u>www.deepnext.ai</u>
- The tools listed above will get better and better
- Programmers will still be needed,
 but their work will look different.
- At some stage, we will no longer understand the code.

