

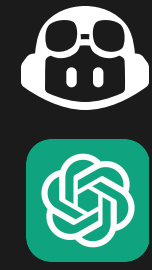
# Use of Code Generation Models in Programming Education: a Systematic Literature Review

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## 1 Background

AI-driven code generation models offer opportunities and challenges for both **educators** and **students** in programming education, such as:



- + they allow students to concentrate on problem-solving components by correcting syntax errors [1], novice programmers perform better & faster with less frustration [2]
- + they help educators create curricula by generating programming exercises and solution explanations[1].
- there are concerns regarding academic integrity and users' over-reliance on auto-generated code [2].

These models have the potential to transform how programming is taught and learned. However, there is still a **lack of understanding of how best to adapt our educational practices to effectively manage the challenges and benefits** associated with their use.

## 2 The Research Questions

*How can code generation models be used in practices for teaching and learning programming?*

- **RQ1:** What are the practices that use code generation models for teaching and learning programming?
- **RQ2:** What are the characteristics of the code generation models that are used in teaching and learning practices?
- **RQ3:** What indicators are used for evaluating the performance of code generation models in teaching and learning practices?
- **RQ4:** What aspects should be considered when utilizing code generation models in teaching and learning practices?

## References

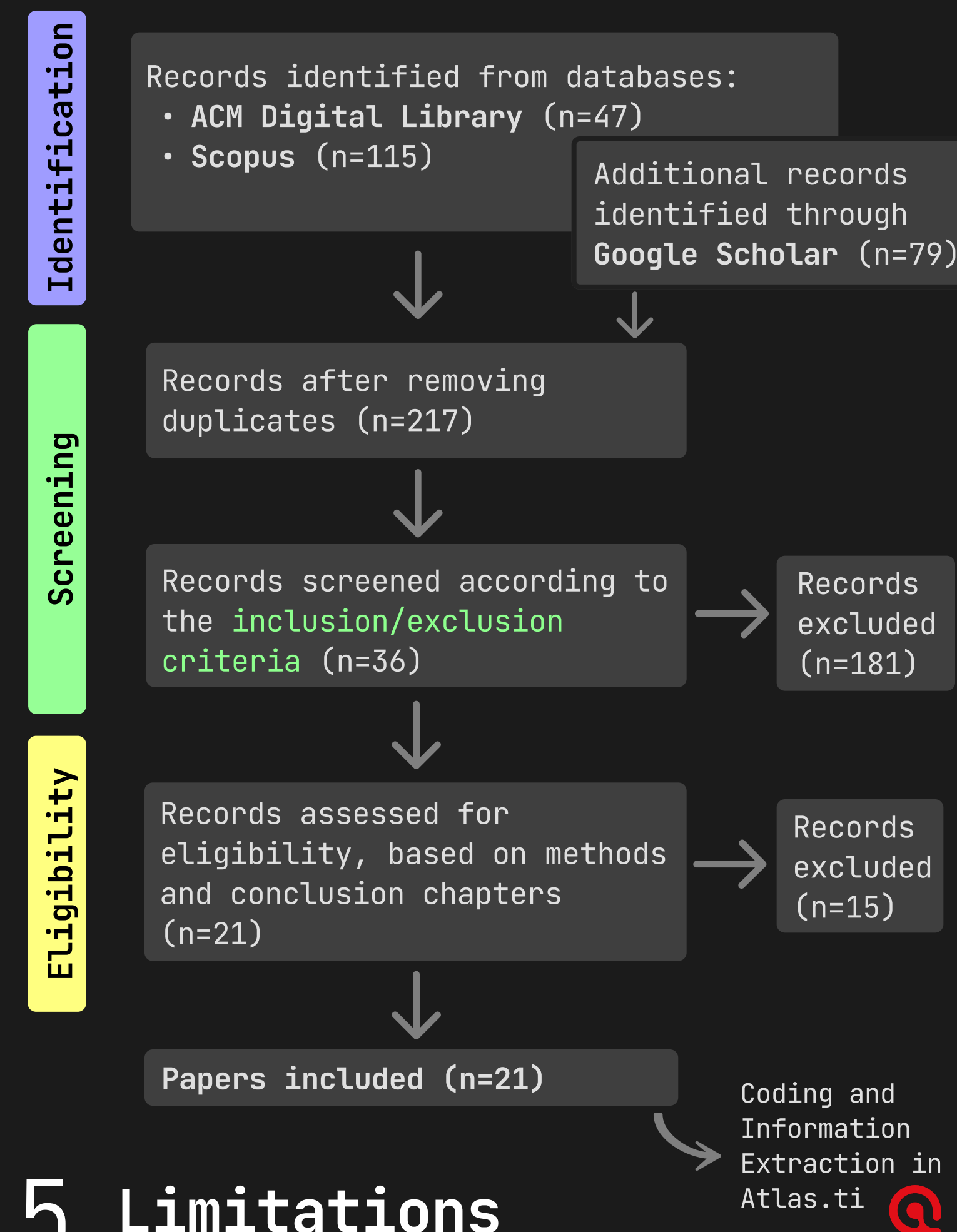
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## 3 Methodology

### Systematic Literature Review

- to identify all empirical evidence that fits the pre-specified criteria and minimize research bias

**PRISMA Flow Diagram:** the paper selection process



## 5 Limitations

- lack of solid guidelines
- the dynamic nature of large language models
- studies only focus on programming education in English

## 4 Results



### Teaching Programming: Assistive tools for assessment generation and evaluation

- Automatic generation of programming assignments, sample answers and explanation, test cases, variations of questions
- Grade Assignments, generate feedback, identify areas students are struggling

### Learning Programming: Virtual Tutors for learners



- Generate practise exercises, exemplar solutions and solution alternatives
- Explain and improve student code, clarify error messages and provide suggestions, support conceptual understanding, provide syntax tips



### Evaluating the Code Generation Models

- evaluating generated content according to **qualitative criteria** (sensibleness, novelty, topicality, readiness for use) and **quantitative criteria** (accuracy, how many tests passed, how many lines of code is explained, etc.) [3,4]

### Challenges and Ethical Considerations

- Academic Integrity
- Over-reliance on the tools, leading to loss of creativity and critical thinking
- Appropriateness to novice programmers
- Accuracy and Reliability issues of the models
- Harmful biases in AI
- Code reuse and licensing issues

### What to do?

- embrace and integrate the code generation tools instead of focusing on detecting and preventing their use
- leave a transitional period for novice programmers



## 6 Conclusion

- Code generators in programming education presents a promising avenue with possibilities to improve student's learning experience and alleviate the workload of teachers.
- However, ensuring safe usage is crucial as failure to address the models' accuracy limitations, risk of misconduct and over-reliance pose a significant danger to computing education.
- Future studies should explore integrating AI code generators in classrooms and designing programming assessments that encourage critical thinking rather than relying on these tools as answer generators.

