# Self-correctors

Key idea: plug in a learned corrector that iteratively improves outputs.



#### Self-corrector = generator + learned corrector

Self-correctors offer several benefits, including:

- . **Controlling** generators without modifying them
- . **Decomposing** problems into multiple iterations
- 3. Using **natural language feedback** for (1) and (2)

## Learning without annotated corrections



Figure 1. Self-corrective learning iteratively trains a corrector by generating hypotheses and corrections, forming value-improving pairs, and selecting those with high similarity for learning.

Generating Sequences by Learning to [Self-]Correct

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## How do we **control** and **improve** a language model's generations **after it is trained**?



## Self-correction

Self-correctors improve upon the base generator, and natural language feedback brings additional gains. Diverse tasks: toxicity, lexical constraints, mathematical program synthesis.



Feedback sources:

- **Toxicity**: fine-grained properties from Perspective API, e.g.
- Lexical constraints: missing words, e.g. add 'bow' and 'prepare'
- Math: few-shot prompted GPT-3, e.g. 2 is missing

### **Correcting inaccessible larger models & multiple corrections**



- Natural language feedback: sources (e.g., humans, models) and formats (e.g., line-by-line).
- Other **learning algorithms** for the corrector: e.g. reinforcement learning

# (Mishra et al., 2022)