Feedback-Driven Semi-Supervised Synthesis of Program Transformations

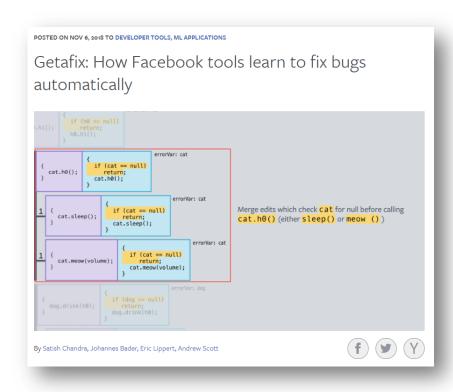
Xiang Gao, Shraddha Barke, Arjun Radhakrishna, Gustavo Soares, Sumit Gulwani, Alan Leung, Nachiappan Nagappan, Ashish Tiwari



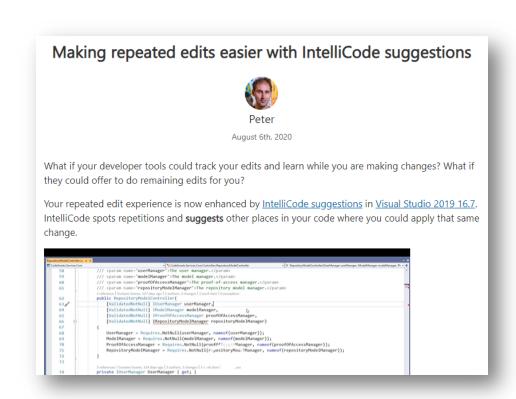




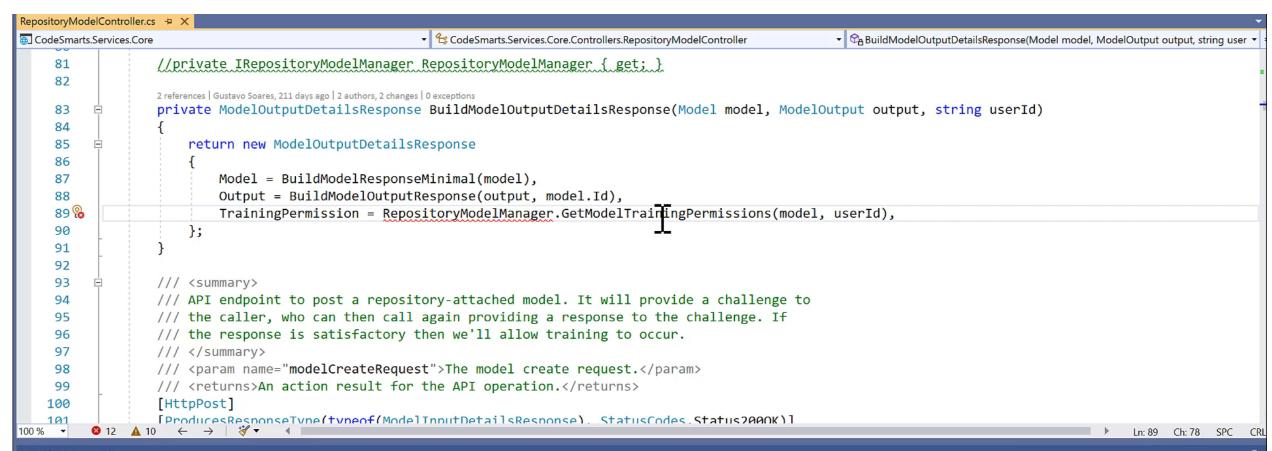
Learning Program transformations from examples



Bader et al. [OOPSLA'19]



Milter et al. [OOPSLA'19]

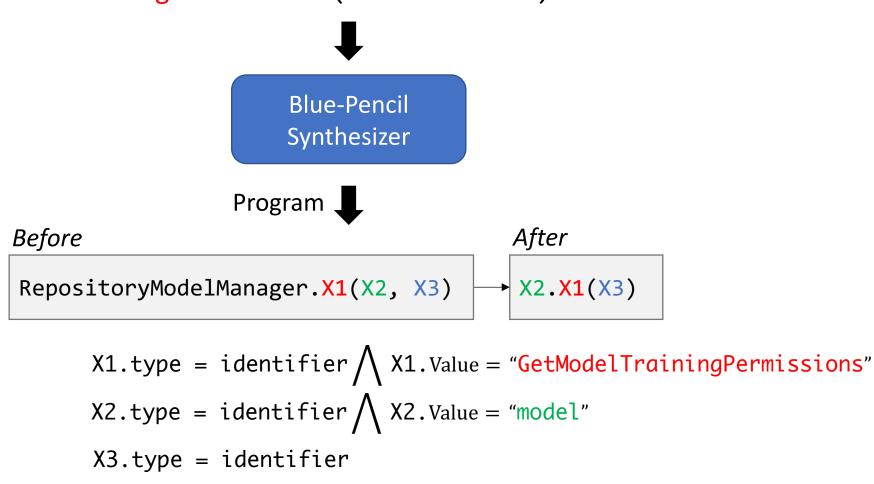


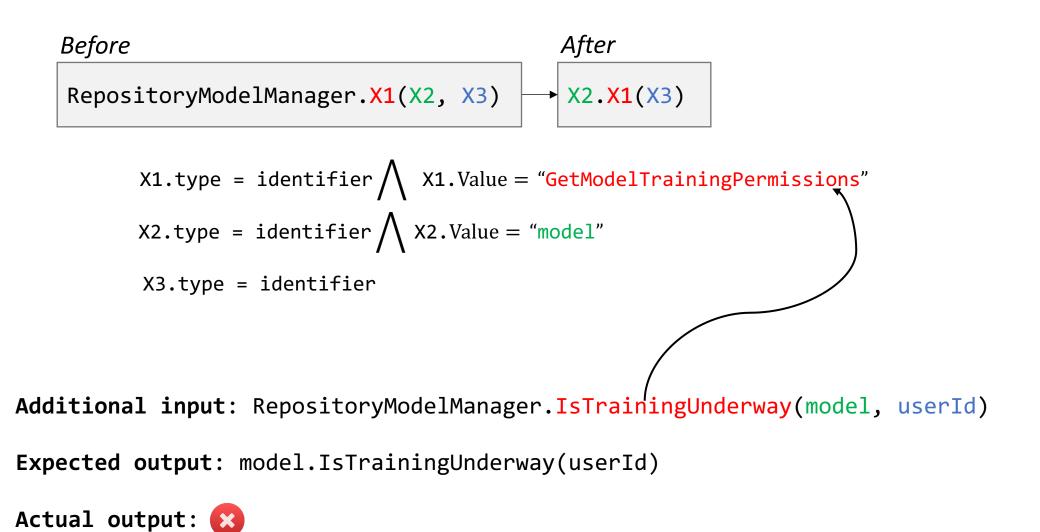
IntelliCode suggestions

Suggestions based on recent edits

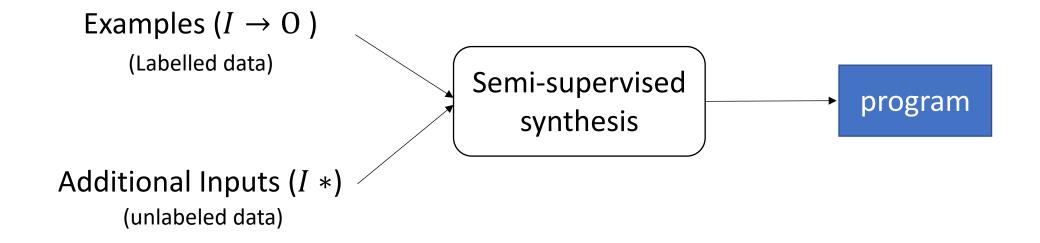


- RepositoryModelManager.GetModelTrainingPermissions(model, userId)
- + model.GetModelTrainingPermissions(userId)
- RepositoryModelManager.GetModelTrainingPermissions(model, currentUserId)
- + model.GetModelTrainingPermissions(currentUserId)

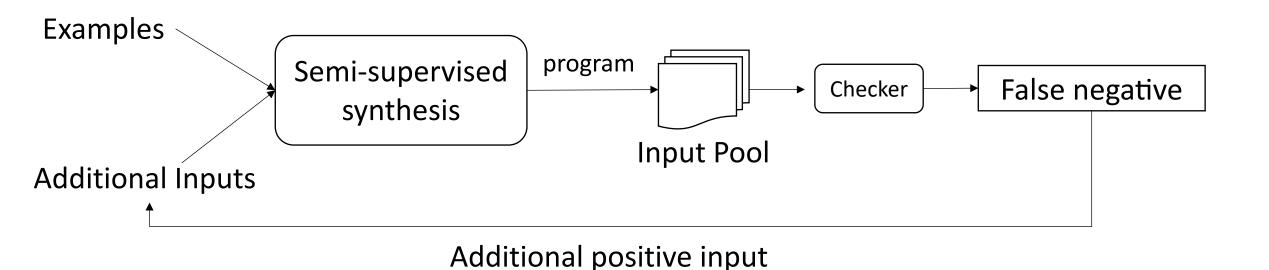




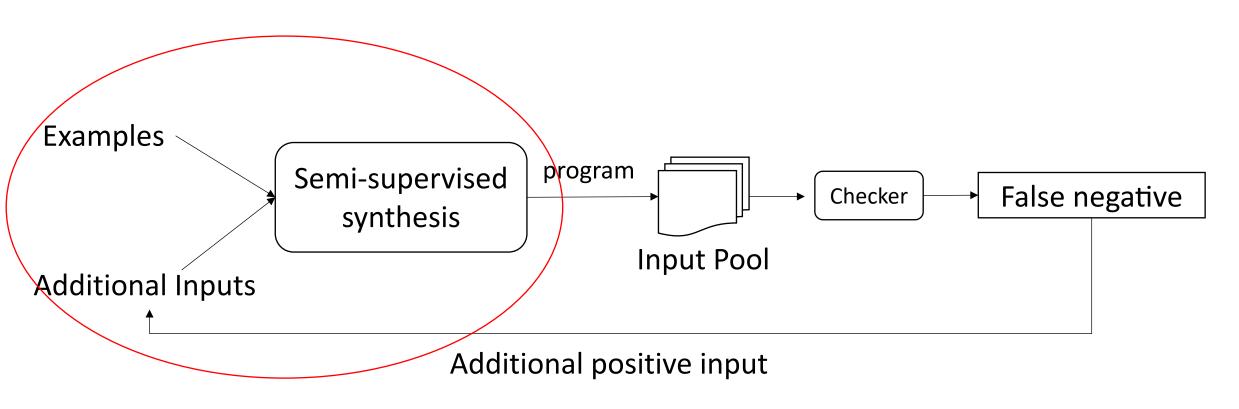
Semi-supervised synthesis of program transformations



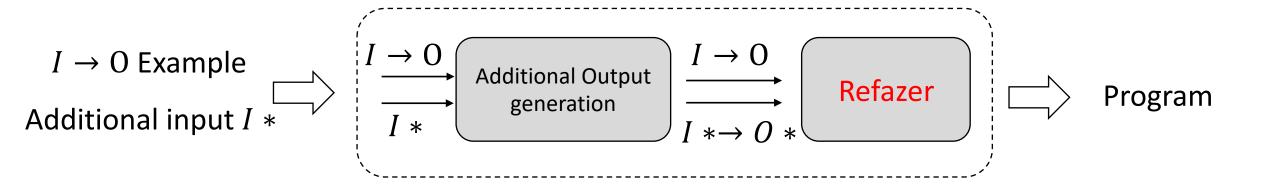
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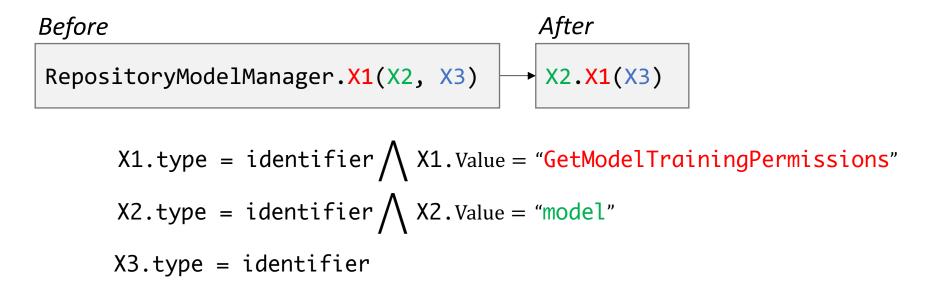


Semi-supervised synthesis



Provenance analysis

Calculate a set of maps between input and output.



Before: RepositoryModelManager.GetModelTrainingPermissions(model, currentUserId)



After: model.GetModelTrainingPermissions(currentUserId);

Anti-unification modulo provenance

input: RepositoryModelManager GetModelTrainingPermissions(model, currentUserId)

additional input: RepositoryModelManager(.IsTrainingUnderway(model,(userId))



Anti-unification modulo provenance

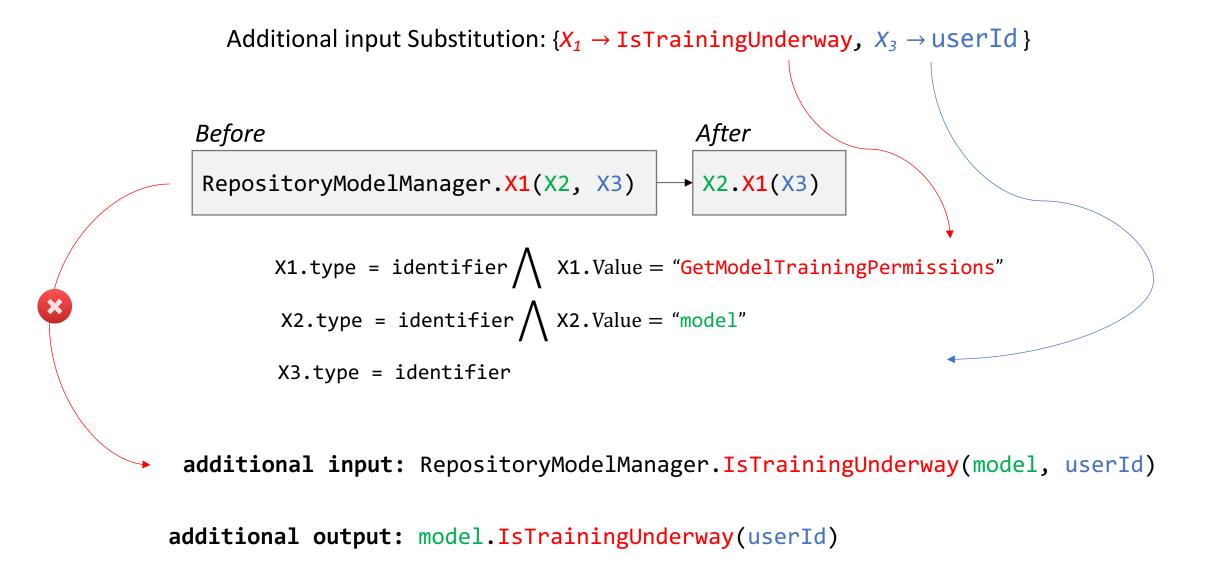


Generalization:

RepositoryModelManager X_1 (model, X_3)

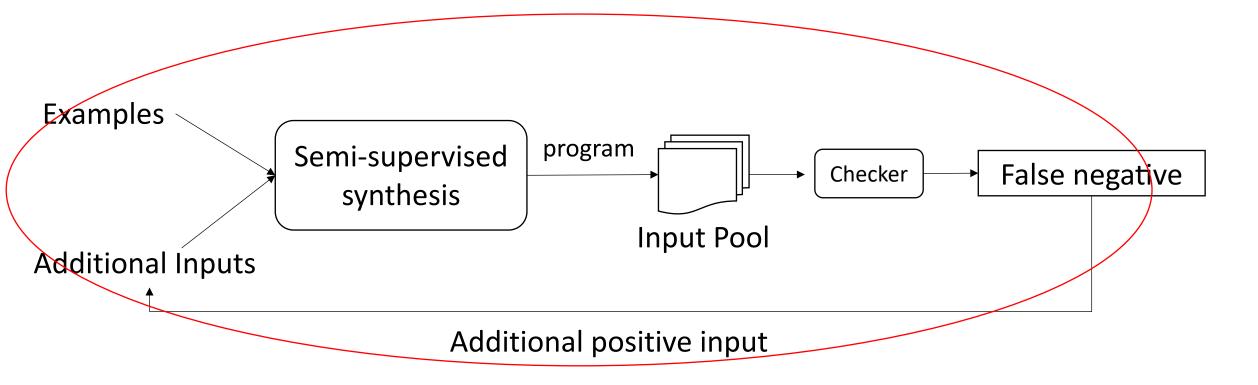
Input Substitution: $\{X_1 \rightarrow \text{GetModelTrainingPermissions }, X_3 \rightarrow \text{currentUserId}\}$

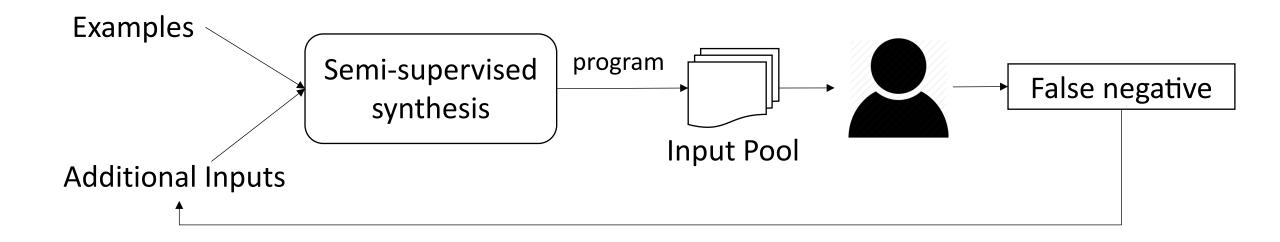
Additional input Substitution: $\{X_1 \rightarrow IsTrainingUnderway, X_3 \rightarrow userId\}$



Example:
$$I \to 0$$
 Refazer Program Additional example: $I * \to 0 *$

Feedback Driven Semi-supervised synthesis of program transformations



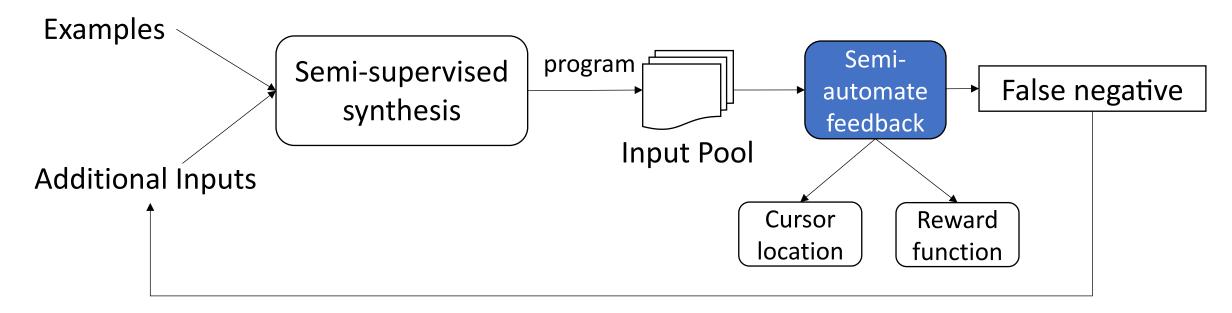


Input pool

```
var userId = HttpContext.GetCurrentUserId();
if (string.IsNullOrWhiteSpace(currentUserId))
{
   return Unauthorized();
}

if (string.IsNullOrWhiteSpace(modelCreateRequest.RepositoryId))
{
   return BadRequest();
}

RepositoryModelManager.IsTrainingUnderway(model, userId)
...
```



```
if (string.IsNullOrWhiteSpace(currentUserId))
{
   return Unauthorized();
}

if (string.IsNullOrWhiteSpace(modelCreateRequest.RepositoryId))
{
   return BadRequest();
}
RepositoryModelManager.IsTrainingUnderway(model, userId);
```

RepositoryManage model.IsTrainingUnderway
RepositoryManager
RepositoryManager

Evaluation

• What is the effectiveness of semi-supervised synthesis in generating correct code transformations?

• Given a cursor location, what is the effectiveness of semi-automated feedback?

Evaluation - semi-supervised synthesis

Scenario	W/O additional input		With additional input	
	Recall	Precision	Recall	Precision
1400	26.71%	100%	100%	96.01%
			5)	5

Semi-supervised synthesis significantly improves the recall of baseline from 27% to 100% while retaining the high precision in generating correct suggestions.

Evaluation - semi-automated feedback

Scenarios	Suggestion	False Positive	False Negative
295	291	1	3

Given a few past edits and one cursor location, our approach achieves around 99% precision and recall in generating correct suggestions.

Conclusion



Existing synthesis produces overfitted programs;



Semi-supervised synthesis;



Feedback loop to generate additional inputs;



We implemented our idea as a VS plugin;



Evaluation results show we could improve the recall without loss too much precision.



