

Automatic Code Review by Learning the Revision of Source Code

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Code keeps on changing



Developer

I've revised the code, is that ok?

```
...
88 private static String
   listToString(List<String> list) {
89   if (list == null)
90     return "";
91   StringBuilder sb = new
   StringBuilder();
92   for (String item : list) {
93     sb.append(item);
94     sb.append(',');
95   }
96   sb.deleteCharAt(sb.length());
97   return sb.toString();
98 }
99 }
```

before revision

revision

```
...
88 private static String
   listToString(List<String> list) {
89   if (list == null)
90     return "";
91   StringBuilder sb = new
   StringBuilder();
92   for (String item : list) {
93     sb.append(item);
94     sb.append(',');
95   }
96   sb.deleteCharAt(sb.length() - 1);
97   return sb.toString();
98 }
99 }
```

after revision

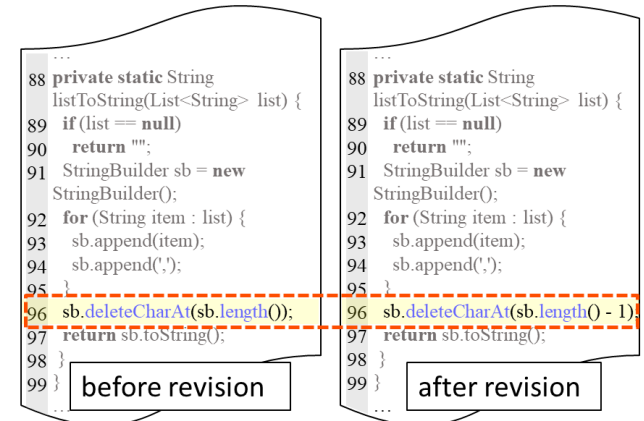
**How to
determine?**

Key point:
Correctly **model
the revision**
between source
code.

Modeling the revision is hard

The state-of-the-art software mining approaches leverage the *global* information by modeling the relationship between two pieces of code.

But apparently, model the revision needs to emphasis the *local* information, i.e., the revised small portion of codes.



Key challenge: model the revision of source code and avoid being misled by the overwhelming amount of unchanged code in the revision.

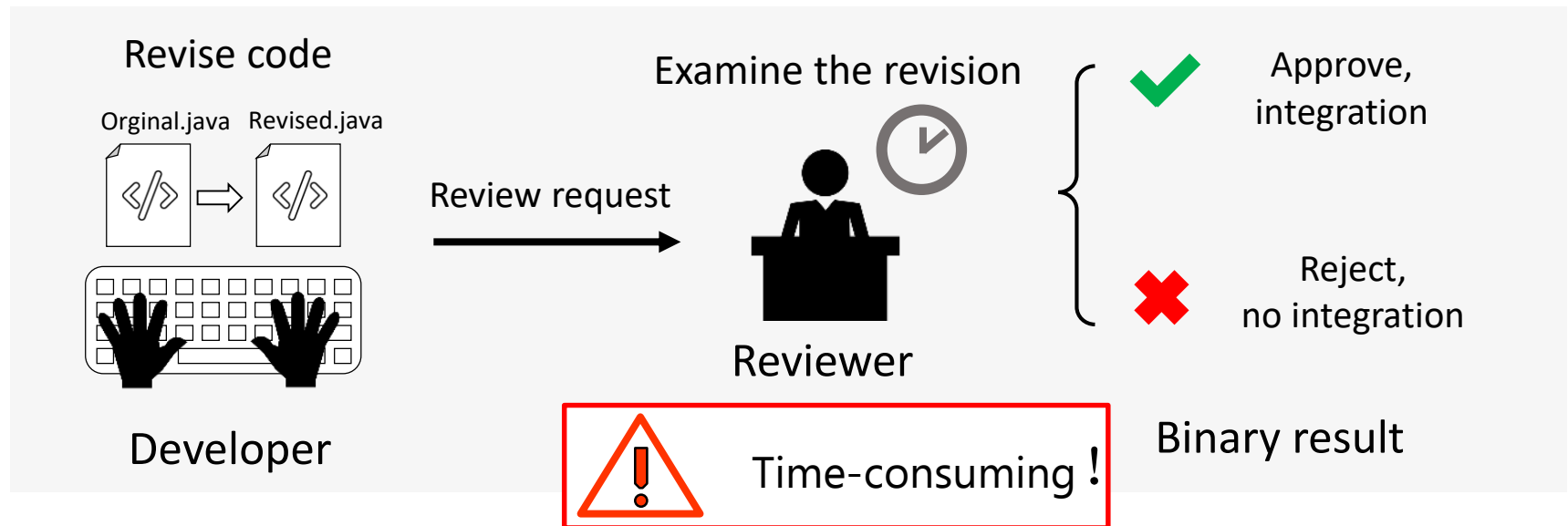
⇒ **This work**

Outline

- What is code review?
- The proposed approach DACE
- Experiments
- Conclusion

Code Review

Code review is the process of inspection on two versions of a source code in order to improve the overall software quality and ensure the security of the project.

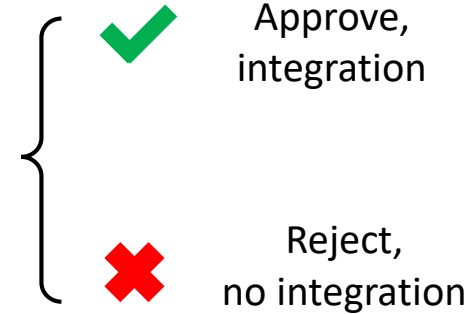
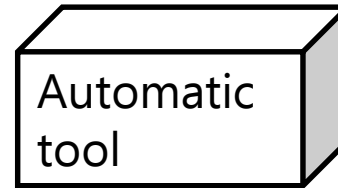


automating such the code review process will alleviate the burden of code reviewers and speed up the software maintenance process.

Automatic Code Review

```
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91   StringBuilder sb = new
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92   for (String item : list) {
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94     sb.append('.');
95   }
96   sb.deleteCharAt(sb.length());
97   return sb.toString();
98 }
99 }
before
```

```
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88 private static String
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95   }
96   sb.deleteCharAt(sb.length() - 1);
97   return sb.toString();
98 }
99 }
after
```



Key challenge: how to model the revision.

Existing models for modeling the relationship between software artifacts

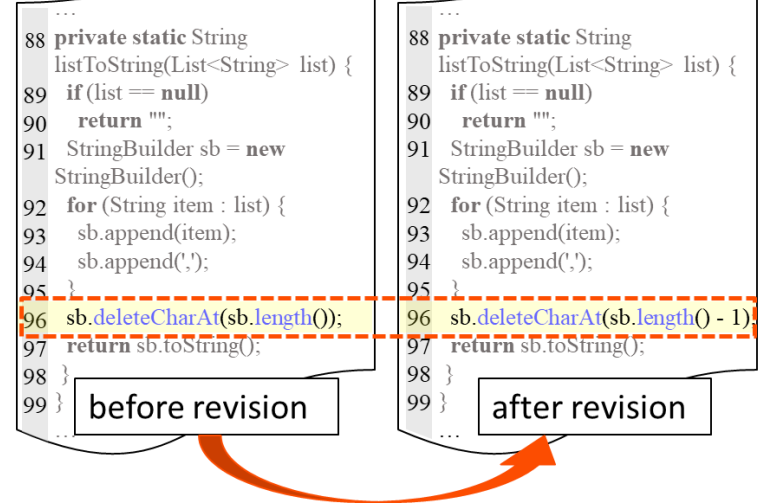
- NPCNN [Huo and Li, 2017]
- CDLH [Wei and Li, 2017]
- LSCNN [Huo and Li, 2018]
- CDPU [Wei and Li, 2018]

*these approaches are equipped for extract **global** feature to model the "similarity".*

But none of them can focus on the **local** feature to model the "**difference**".

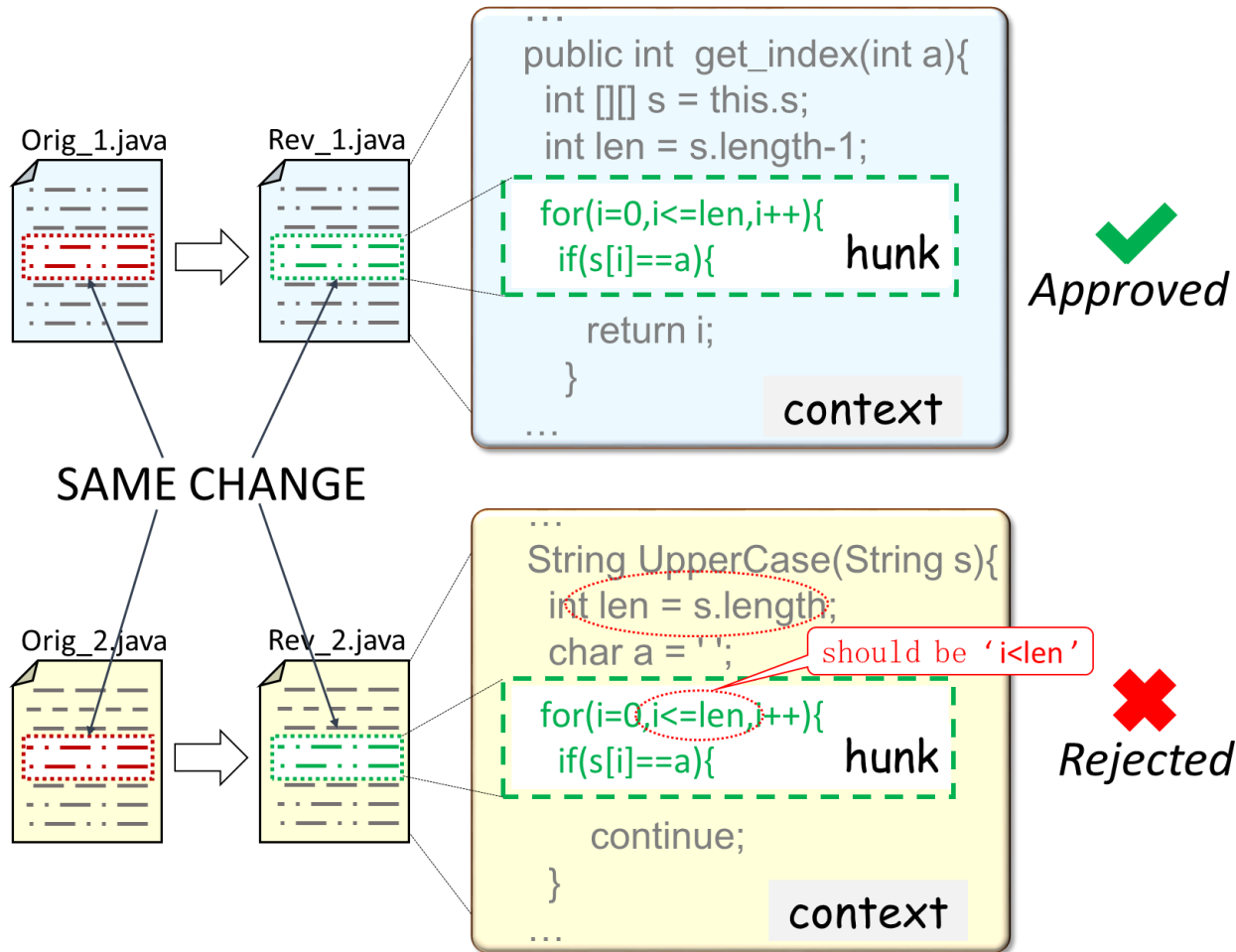
Challenge 1

**How to model the revision
of source code?**

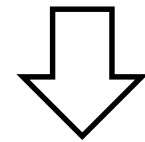


Key idea: A particularly designed pairwise autoencoder (PAE) is employed to learn the revision features. We construct an encoder to pack the revised code into a compact feature representations based on which the transformation can be learned.

Challenge 2



Same change
different result.



Context of the
changed hunks
provides rich
information about
the revision.

Challenge 2

- The changed hunk is usually *short*, while the context is usually *long*.
- **Dilemma**
 - Utilize all the context ?
overwhelmed by same unchanged code lines. ✖
 - Discard the context ?
the revised code is too short to provide sufficient information for modeling . ✖

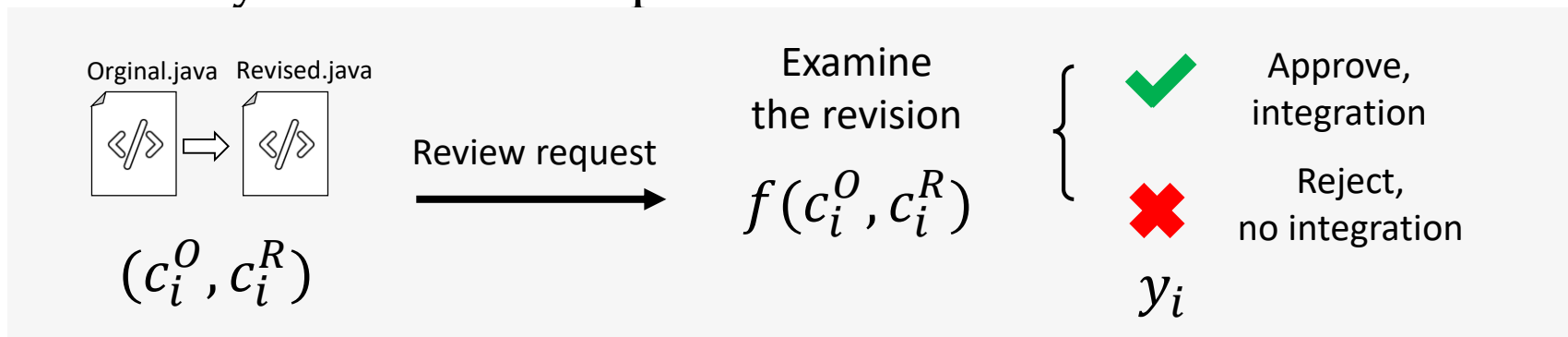
How to **focus on modeling the revision but also considering the rich information in the context?**

Key idea: Enrich the revised lines with the information of context around them.

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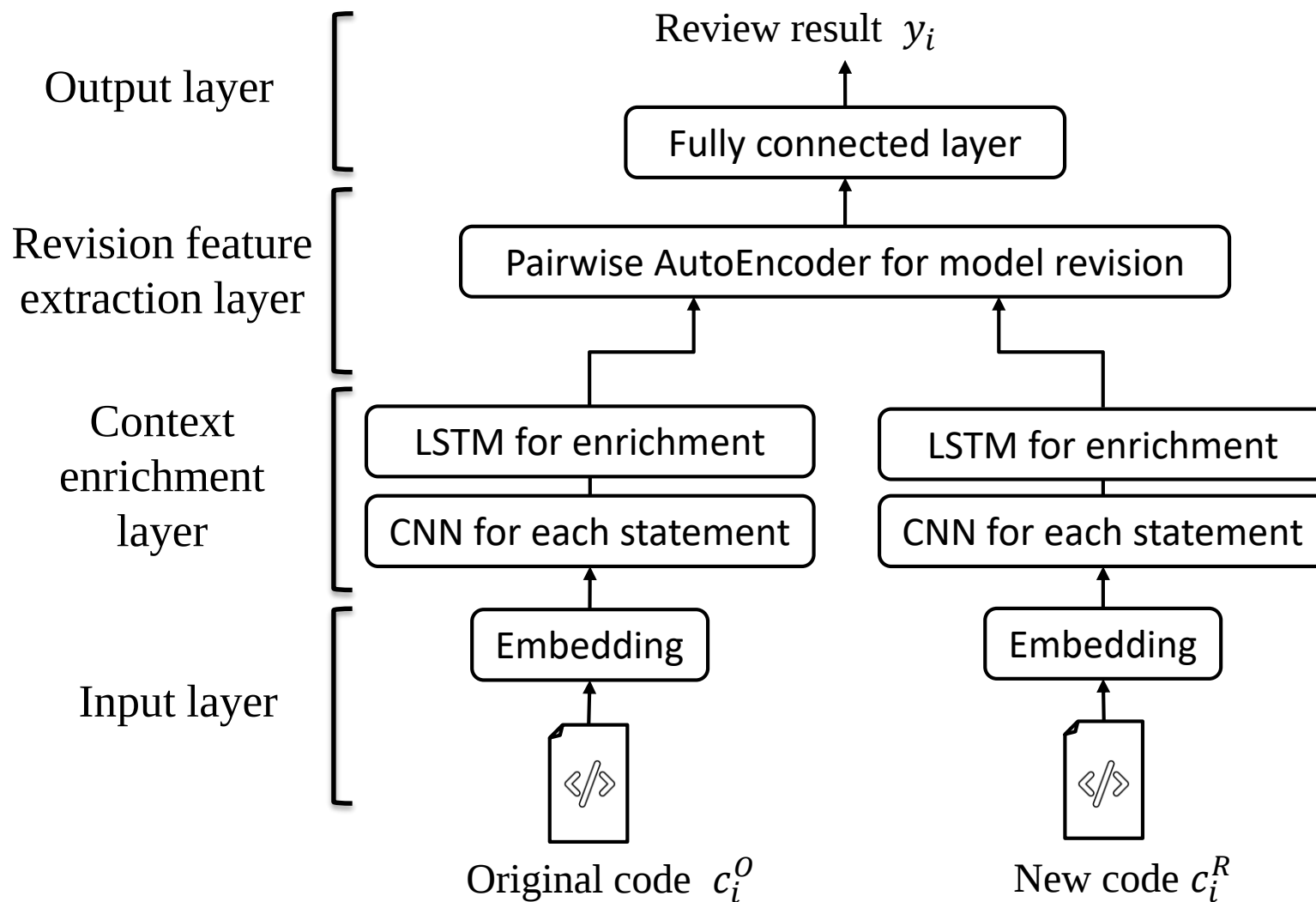
- We formalize the code review as a learning task, which is a binary classification problem.



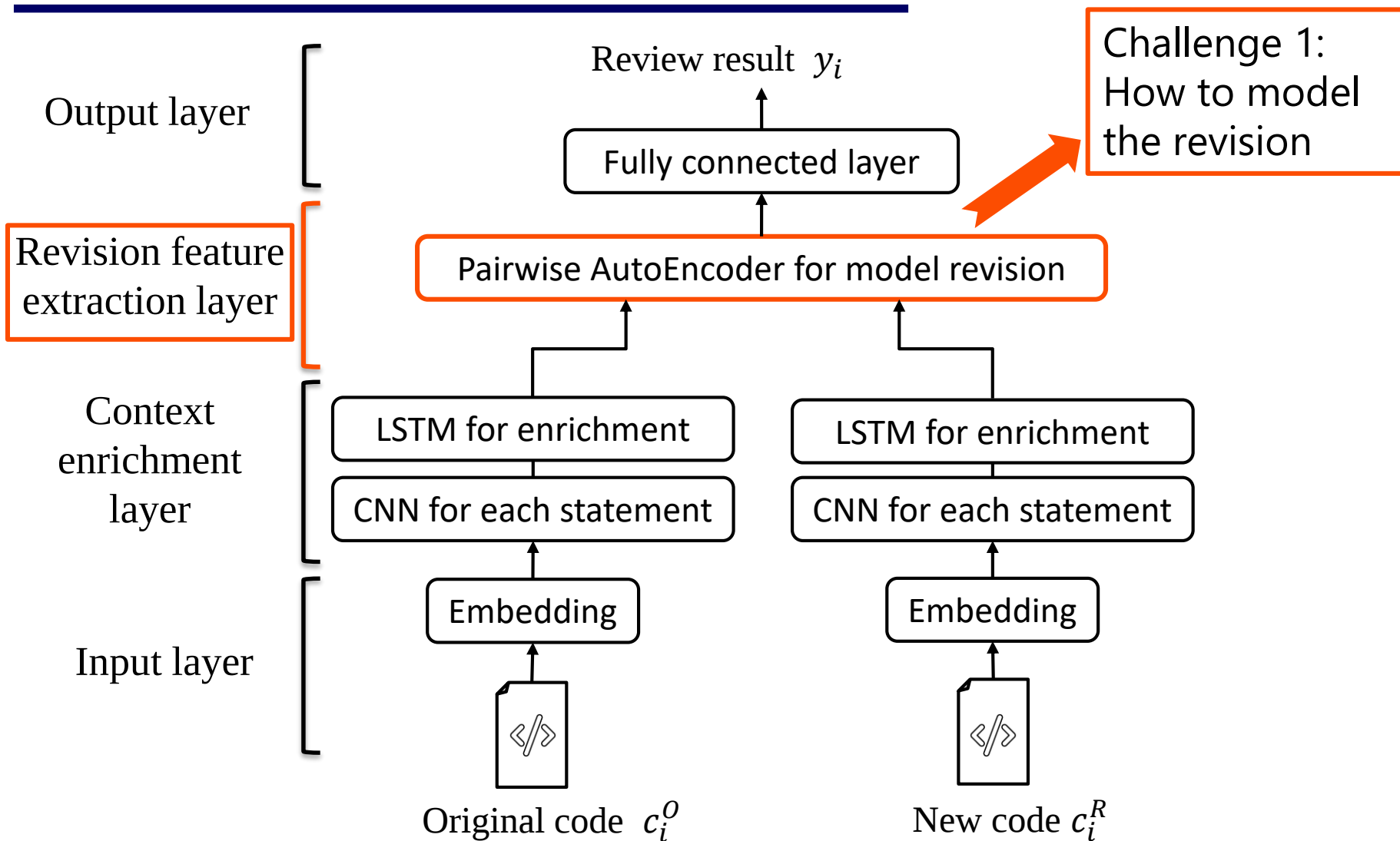
- Given a sample of data: (c_i^O, c_i^R, y_i) , $c_i^O \in \mathcal{C}^O, c_i^R \in \mathcal{C}^R$
- Learn a mapping $f: \mathcal{C}^O \times \mathcal{C}^R \mapsto \mathcal{Y}$ that can be learned by minimizing the objective function:

$$\min_f \sum_i \mathcal{L}(f(c_i^O, c_i^R), y_i) + \lambda \Omega(f)$$

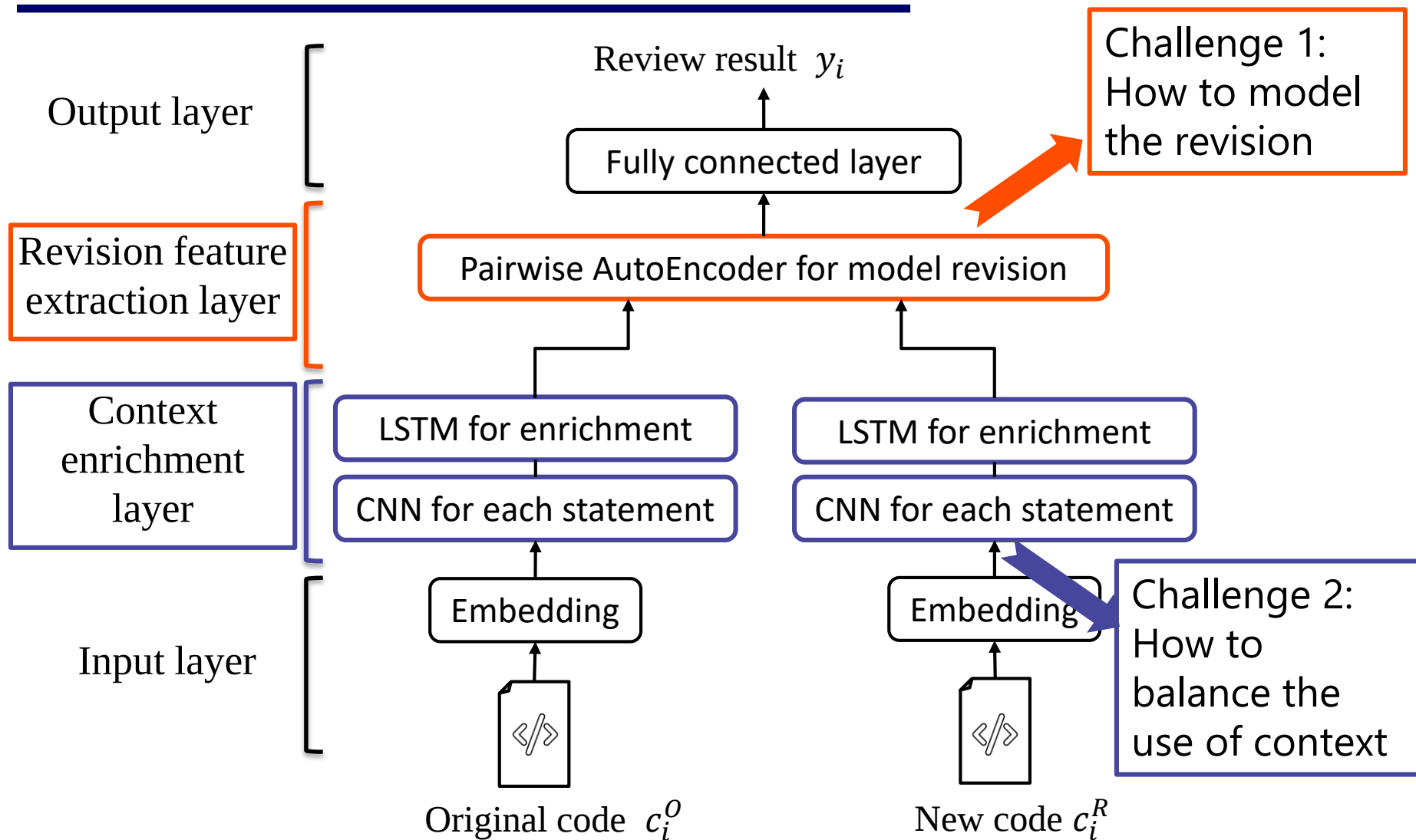
Model Structure of DACE



Model Structure

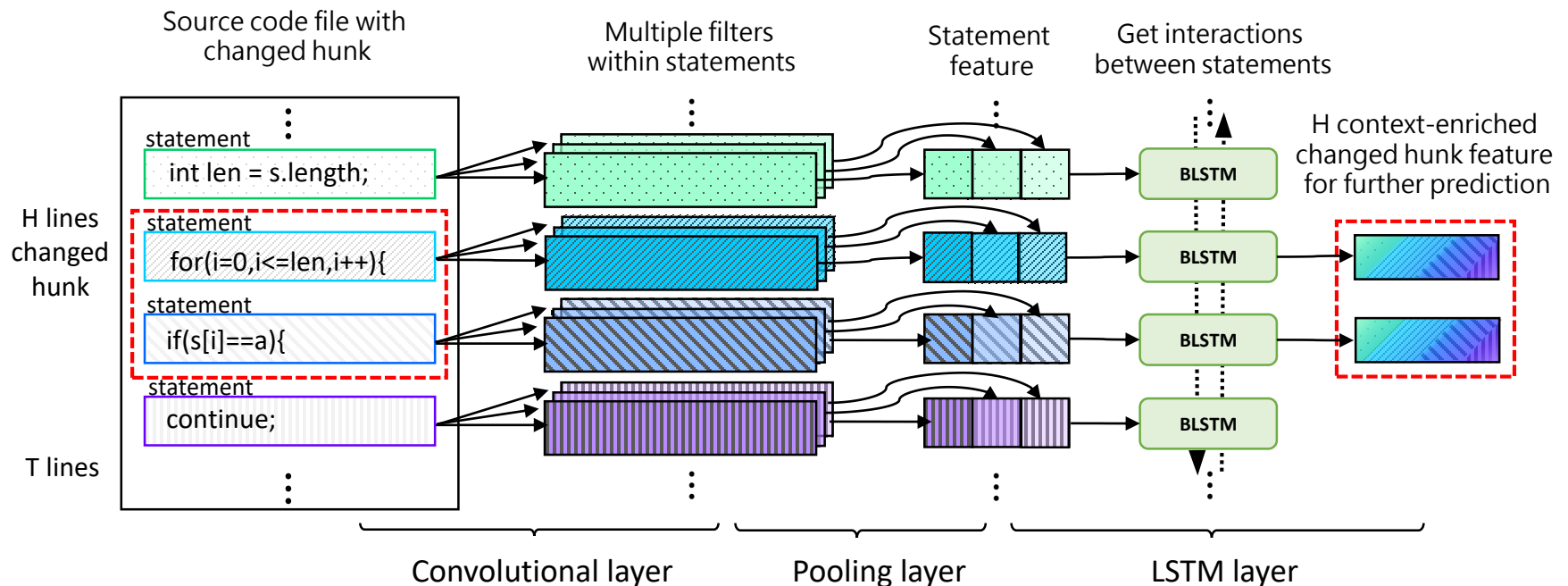


Model Structure

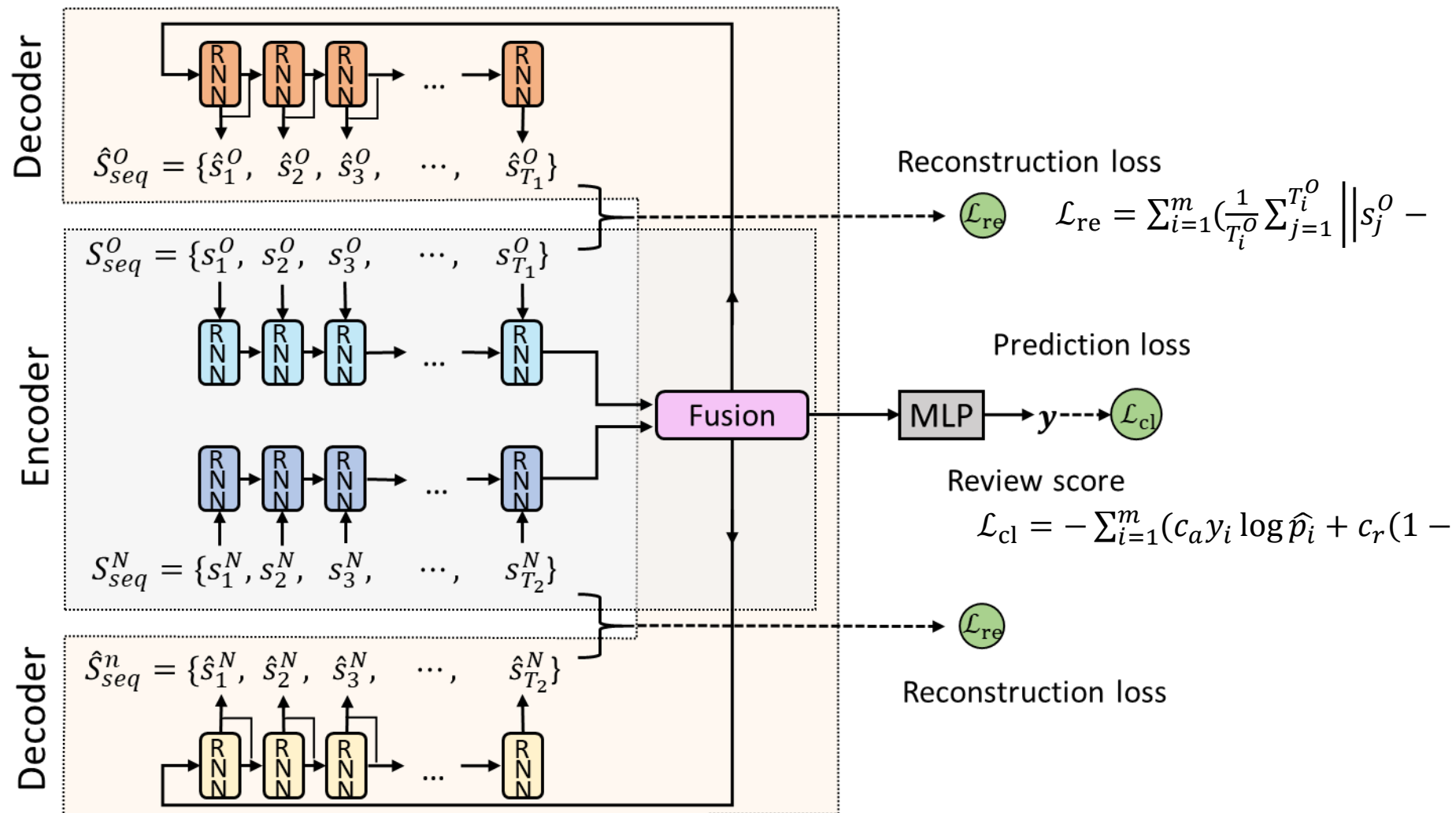


Context Enrichment Layer

- Extract the statement level features by convolutional layer
- Enrich the current statement with other statements in the context by exploiting the sequential relationship by LSTM layer



Revision Feature Extraction Layer



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Experiment Setting

- Dataset

- The code review data used in the experiments is crawled from Apache Code Review Board.
- For each changed hunk, if it has the highlighted lines that marked by reviewers denoting they have issues, we regard the hunk as rejected.

- Evaluation measure:

- ✓ AUC
- ✓ F1

Repository	#hunk
accumulo	5,620
ambari	6,810
aurora	6,762
cloudstack	6,171
drill-git	3,575
hbase-git	6,702

Experiment

- **TFIDF-LR** : uses TFIDF to represent the original and the revised source code, and LR is used for prediction.
[Schütze *et al.*, 2008][Gay *et al.*, 2009]
- **TFIDF-SVM**: like above but SVM is used for prediction.
[Schütze *et al.*, 2008][Gay *et al.*, 2009]

Non-deep
methods

- **Deeper** : a state-of-the-art model on software engineering, which uses basic features for changes.
[Yang *et al.*, 2015]
- **Deeper-SVM**: like above but SVM is used for prediction.
[Yang *et al.*, 2015]

Deep methods

- **LSCNN**: a variant of DACE that without PAE. It is similar to the state-of-art method LSCNN and using Multilayer Perceptron (MLP) for prediction.
- **PAE** : a variant of DACE that only use PAE to extract features from CNN without considering sequential and long-term dependency information between statements.

Variants

Experiment

Repository	TFIDF-LR		TFIDF-SVM		Deeper		Deeper-SVM		LSCNN		PAE		DACE	
	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC
accumulo	0.227	0.666	0.239	0.703	0.202	0.688	0.199	0.705	0.417	0.787	0.373	0.814	0.493	0.786
ambari	0.240	0.708	0.278	0.848	0.306	0.680	0.238	0.572	0.444	0.824	0.473	0.861	0.509	0.905
aurora	0.204	0.582	0.220	0.645	0.349	0.682	0.299	0.564	0.336	0.750	0.571	0.819	0.403	0.793
cloudstack	0.250	0.745	0.275	0.827	0.352	0.795	0.265	0.646	0.360	0.761	0.415	0.820	0.516	0.852
drill-git	0.212	0.658	0.236	0.725	0.229	0.593	0.212	0.540	0.318	0.788	0.382	0.806	0.573	0.820
hbase-git	0.232	0.679	0.256	0.759	0.193	0.590	0.154	0.524	0.348	0.751	0.411	0.764	0.396	0.813
average	0.228	0.673	0.251	0.751	0.272	0.671	0.228	0.592	0.370	0.777	0.438	0.814	0.482	0.828

DACE outperforms other state-of-the-art competing methods in term of F1 and AUC.

Experiment

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	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC	F1	AUC
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The internal structure of pairwise autoencoder helps DACE to captures the revision by learning the revision patten by PAE.

Experiment

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average	0.228	0.673	0.251	0.751	0.272	0.671	0.228	0.592	0.370	0.777	0.438	0.814	0.482	0.828

PAE only not perform better than DACE because it abandons the context of revised hunk which is indispensable.

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Conclusion

- Major Contribution
 - Put forward a new software mining problem that aims to **model the relationship between two source files by learning the revision features.**
 - Propose a novel model DACE, which learns the revision features **by modeling the hidden state, based on pairs of context-enriched representation of source code.**
- Future work
 - Considering the relationship between hunks.
 - Incorporating additional data.
 - ...

Thanks!