

# Dec 2 - Dec 9 Progress

Raise an issue.

## Reproduction

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- Can now reproduce Lin Tan's results on PROMISE data. We have his PROMISE and AST baselines, but not his semantic features (the code to generate those are protected by a patent).
- Difference was in how `bug > 1` samples were being treated; they are now set to 1 instead of being purged.

## Experiments

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- Running models with the new pre-processing on PROMISE data, they're doing better.

Results so far:

Dataset	Our method				MSR			Lin Tan		
	P	R	F	Runtime	P	R	F	P	R	F
ant 1.5 - 1.6	51.9 (3.7)	72.8 (10.9)	<b>59.9</b> <b>(1.9)</b>	20.6 (0.4)	33	80	47	44.8	51.1	47.7
ant 1.6 - 1.7	45.6 (2.7)	67.5 (3.0)	<b>54.6</b> <b>(0.7)</b>	37.7 (0.3)	21	98	35	41.8	77.1	54.2
camel 1.2 - 1.4	25.6 (1.3)	55.2 (4.8)	34.6 (1.4)	8.6 (0.8)	20	82	32	24.8	75.2	<b>37.3</b>
camel 1.4 - 1.6	28.6 (1.1)	48.9 (5.9)	35.9 (1.6)	31.5 (0.2)	28	68	<b>40</b>	28.3	63.7	39.1
ivy 1.4 - 2.0	20.4 (2.7)	67.5 (10.0)	32.3 (3.9)	43.6 (0.8)	11	70	18	22.6	60	<b>32.9</b>
jEdit 3.2 - 4.0	39.4 (1.3)	64.0 (4.0)	49.0 (1.6)	49.6 (0.3)	35	75	47	44.7	73.3	<b>55.6</b>
jEdit 4.0 - 4.1	47.8 (2.7)	64.6 (5.0)	<b>55.2</b> <b>(1.4)</b>	14.8 (0.5)	33	93	49	46.1	67.1	54.6
xerces 1.2 - 1.3	16.1 (1.0)	75.4 (17.4)	<b>31.3</b> <b>(3.0)</b>	43.9 (0.3)	23	28	26	16	46.4	23.8

- Where we are not SOTA yet, I will run more hyper-parameter search. The above are done with only the top-10 models, rather than top-28.

## Research Questions

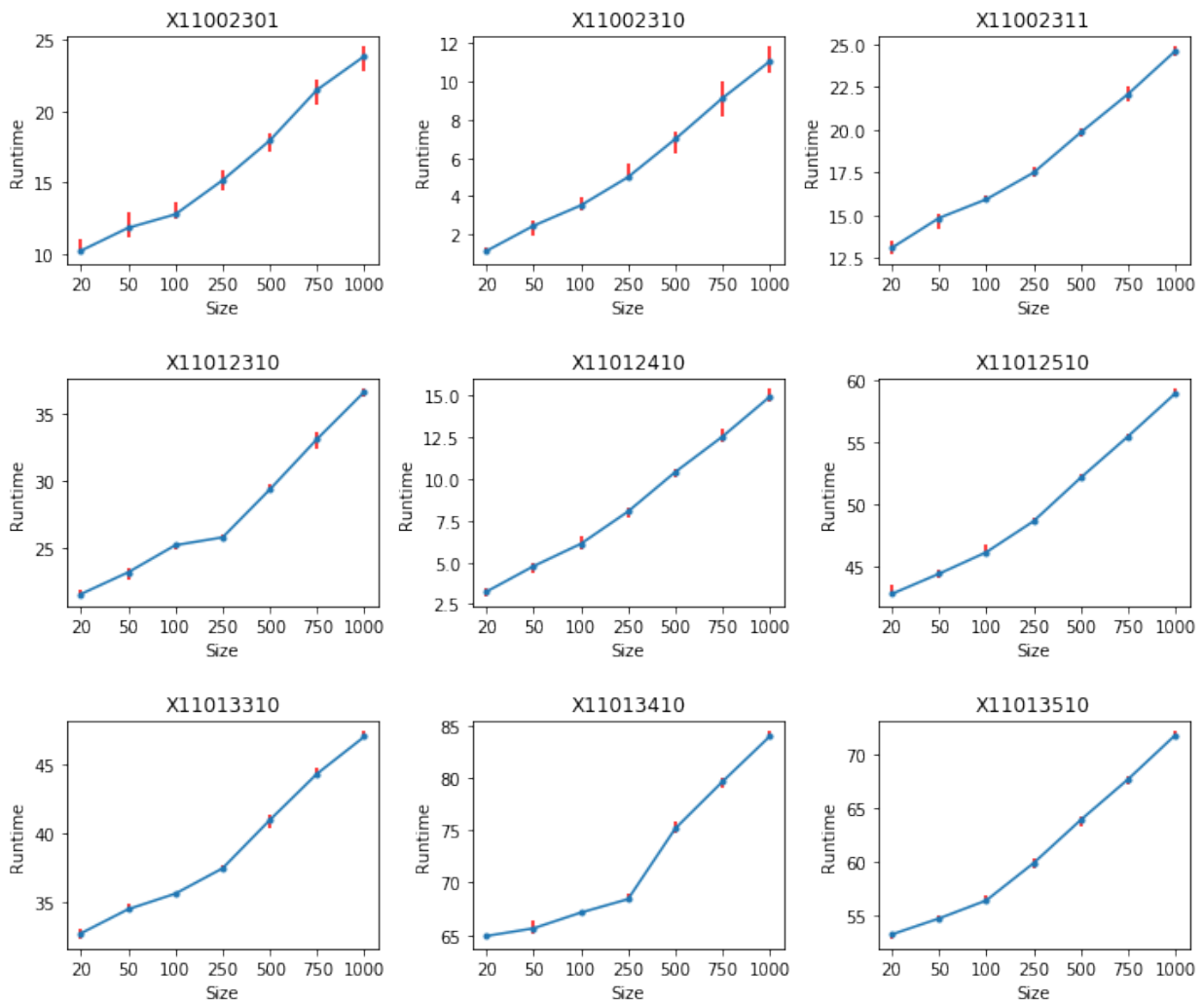
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### Old

- *How far can we stray from DL literature and still do well?*
  - Hamming distance 3
- *Can these models outperform standard DL models?*
  - Yes.
- *Can these models outperform SOTA SE methods?*
  - In general, yes
- *Are these models transferable to other datasets?*
  - Where transfer does NOT mean cross-project defect prediction (CPDP), generally, yes.
  - For CPDP, experiments are required

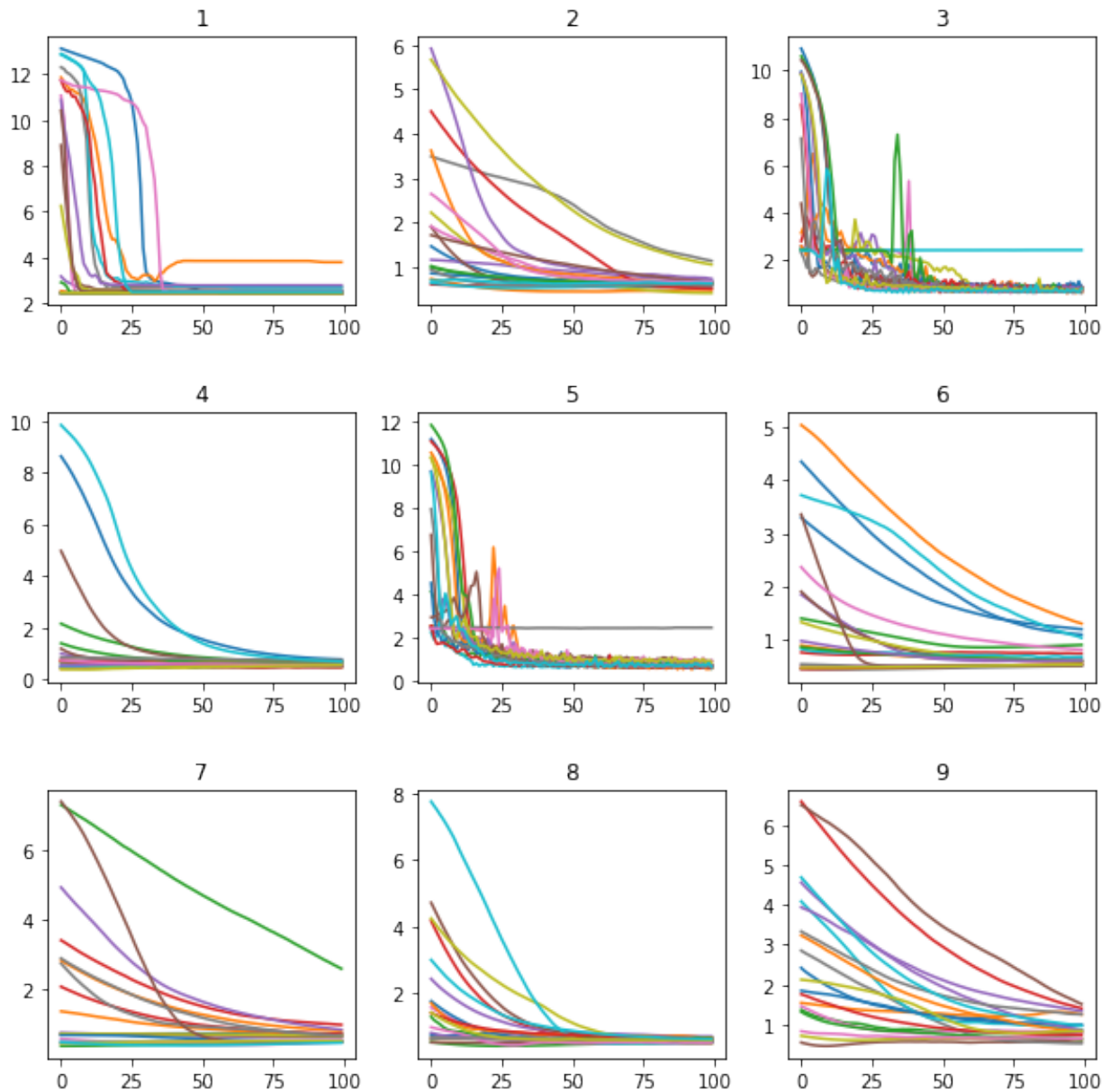
- *Are these models scalable?*

- Yes.



## New

- *Why does oversampling help so much, such that the majority of models choose it?*
  - Because oversampling causes the minority class samples to contribute more to the loss, and therefore, the model cannot ignore them.
  - The performance boost from using these data points (< 5%) comes from oversampling rather than simply using them.
- *How many epochs are really needed to train these models?*
  - About 65-75 seems to work for most models.



## Work for next week

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- Finish experiments
- Deep learners on AST features

## Work for later

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- Run cross-project defect prediction experiments
- Start writing paper (?)

## Other Discussion

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- RA Paperwork?
- Journal to aim for?