CSE Research and Technology Forum 2022





May my Al systems be better assured? Shing-Chi Cheung

Supported by ITC under MHP-055-19

Sound familiar?



A Common Issue in Al Systems

Does my Al system perform reliably?



Defects in AI systems can bring great loss

- Defects occurring at deployment can pose threats to lives and economy
- Defects occurring at training can waste weeks or months of valuable computational resources
 - Result in NaN, crashes and gradient vanishing after a long training period



2018/3/19: Uber car hit a pedestrian, causing death due to incorrect object detection

How frequent are real-life AI projects updated?

Deep Learning Project	Code Size	#Code Commits	#Commits in recent 6 months	#lssues reported in recent 6 month
Transformer	29,409	1,971	1,549	891
DeepSpeech	145,167	2,108	594	252
Real-Time-Voice- Clone	5,126	237	148	148
DeepCreamPy	878	376	86	37

Source: Four most popular active TensorFlow projects on Github Figures based on end of 2019

Can bugs in neural networks be detected before training?



Absence of testing tools for DL systems



There are commercial testing tools for conventional software

BUT none are designed for AI systems or modules

First empirical study on bugs in DL programs

	StackOverflow QAs	Github Projects
Counts	87	88

StackOverflow QAs

- Searched Tensorflow related questions
- Manually reviewed QA pages
- Analyzed answers and discussions

Github Projects

- Searched Tensorflow related projects
- Manually reviewed commits
- Analyzed commit/pull request messages and issue discussions

Yuhao Zhang, Yifan Chen, Shing-Chi Cheung, Yingfei Xiong, Lu Zhang. An Empirical Study on TensorFlow Program Bugs. In ACM SIGSOFT International Symposium on Software Testing and Analysis (ISSTA2018).

Selected findings from popular Github projects

- Failures often occur at the training stage and after many training cycles
- Testing one training instance alone is unlikely to catch such failures
- When failures occur, the error messages are often confusing
 - Error messages may not pinpoint which parts of the software go wrong
 - Discussions at StackOverflow suggest that fault determination is non-trivial even for a small AI system

A common issue in many Al projects

- API misuses are common
 - API mostly designed for numeric computation
 - API documentation is either brief or difficult to follow

h_fc3 = tf.nn.relu(conv2d(h_fc1_drop, W_fc2) + b_fc2)
y_conv = tf.nn.softmax(tf.reshape(h_pool3, [-1, 10]))
cross_entropy = -tf.reduce_sum(y_*tf.log(y_conv))
train_step = tf.train.AdamOptimizer(1e-4).minimize(cross_entropy)

Stack Overflow #33699174

API often evolves to meet dynamic market demand and algorithm advancement

DEBAR: Scalable AI Defect Analyzer

We propose two abstraction techniques:

1. Tensor Partitioning

Numeric computation can be abstracted using intervals Many tensor elements are subject to the same computation

2. Interval Abstraction with Affine Equality Relation

Many computations are affine operations ($w_0 = \sum_i w_i x_i$) -> affine equality relations to enhance the precision of interval abstraction



Distinguished paper award ESEC/FSE 2020

Tool implemented in Python https://github.com/ForeverZyh/DEBAR

Main results

Framework (Tensor Abstraction + Numerical Abstraction)

Our technique

(Tensor Partitioning + Affine Equality Relation): Accuracy: <u>93.0%</u>, all in 3 minutes, 12.1s on average 100% accuracy on 9 buggy architectures

Tensor Partitioning + Sole Interval Abstraction Accuracy: 80.6%, 12.1s on average expand every element $\sigma(A) = \begin{pmatrix} [0,1] & [-1,0] \\ [-1,0] & [1,1] \end{pmatrix}$

> Tensor Expansion + Affine Equality Relation: 33/57 > 30mins; on rest 24, DEBAR doesn't lose accuracy

Tensor Smashing + Affine Equality Relation:

Accuracy: 87.1%, 12.2s on average

Smash a tensor into an element $\sigma(A) = [-2,2]$

Accuracv =

S.C. Cheung@HKUST





embers

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Group

CASTLE





Yepang Liu Asistant Professor at SUSTech



Rongxin Wu Associate Professor at Xiamen University



Ming Wen Associate Professor at HUS





Lili Wei Postdoc at HKUST

CASTLE Group - Code AnalySe, Testing and LEarning



Valerio Terragni

Lecturer at The University of

Auckland

Students

Jue Wang PhD Student at NJU



Huaxun Huang PhD Student at HKUST

Yanyan Jiang

Associate Researcher at NJU



Yonggiang Tian PhD Student at HKUST&UWaterloo



Mijung Kim

Assistant Professor at UNIST

Cong Li PhD Student at NJU



Meiziniu Li PhD Student at HKUST



Lu Liu PhD Student at HKUST





Haoyang Ma PhD Student at HKUST





Jialun Cao PhD Student at HKUST





http://castle.cse.ust.hk/castle/people.html



Hengcheng Zhu MPhil Student at HKUST



Wugi Zhang

PhD Student at HKUST

MPhil Student at HKUST



Associate Professor at NEU

Ying Wang