

Facilitating Postmortem Program Analysis with Deep Learning

Wenbo Guo¹, Xinyu Xing¹, Jimmy Su¹

1. JD Security Research Center





Grim Reality

- Despite intensive in-house software testing, programs inevitably contain defects.
 - Accidentally terminate or crash at post-deployment stage.
 - Exploited as security loopholes.
- At JD, we have several large scale data centers.
 - Contains plenty of software.
 - Report a great many crashes everyday.



Grim Reality

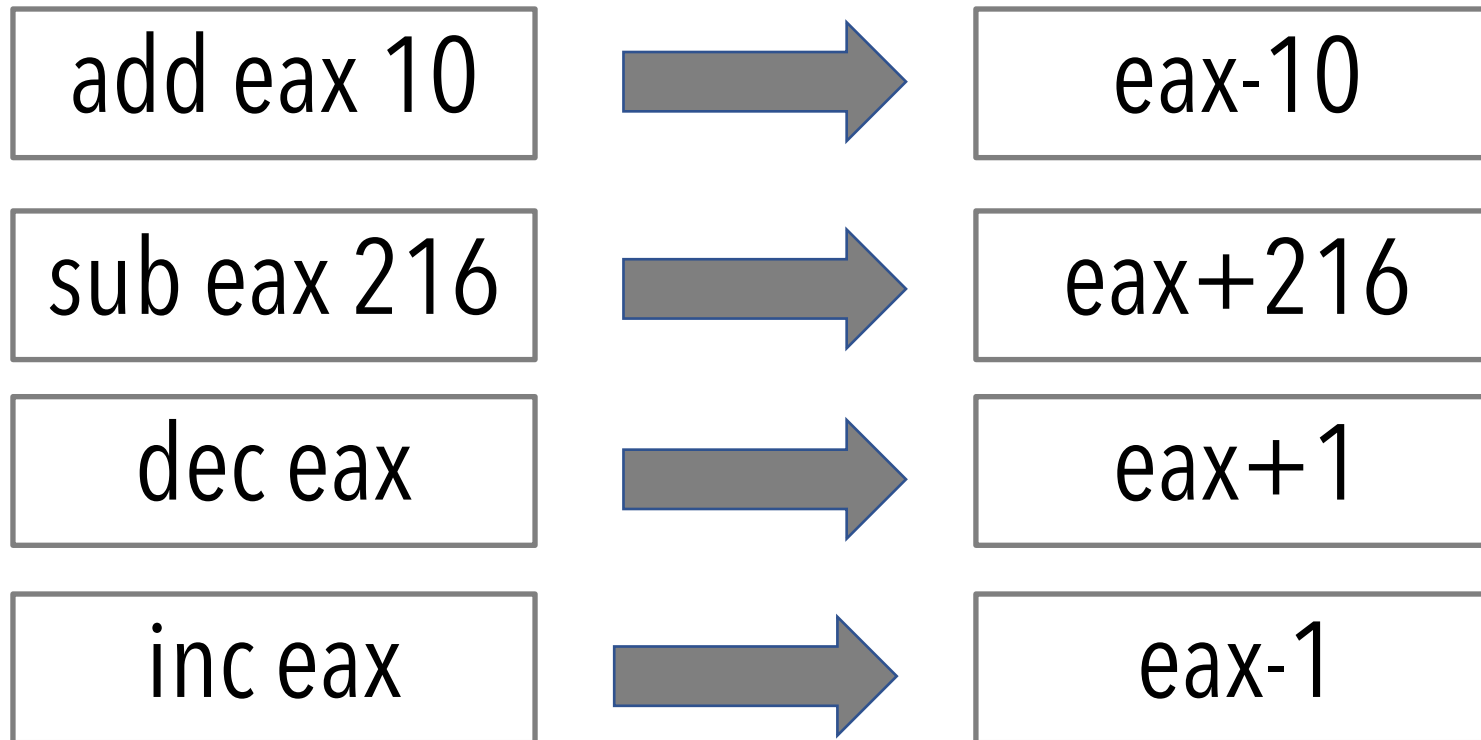
- Software analysts need to debug the program and patch the vulnerability.
- Software debugging is expensive and needs intensive manual efforts.
 - Debugging software cost has risen to \$312 billion per year globally.
 - Developers spend 50% of their programming time finding and fixing bugs.



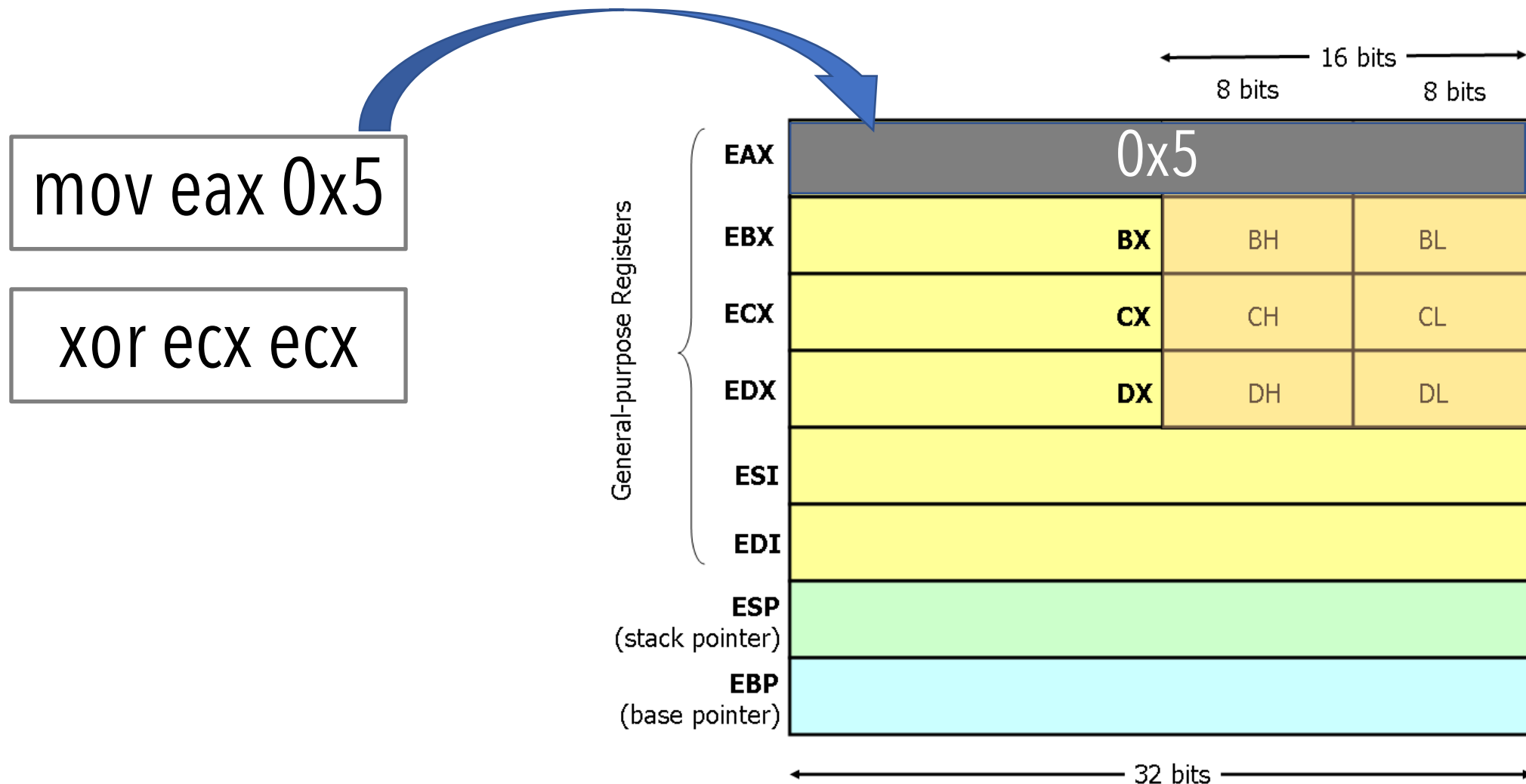
Automatic debugging: Postmortem Program Analysis

- Track down the root cause of a software crash at the binary level without source code.
- Analysis a crash dump and identify the execution path leading to the crash.
 - Reversely execute an instruction trace (starting from the crash site).

Reverse Execution – Invertible Instructions

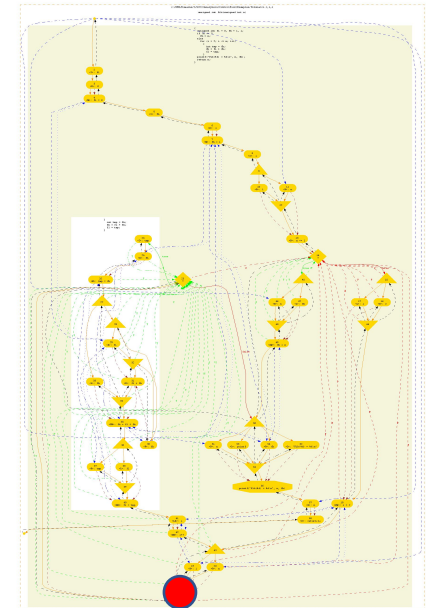


Reverse Execution – Non-invertible Instructions



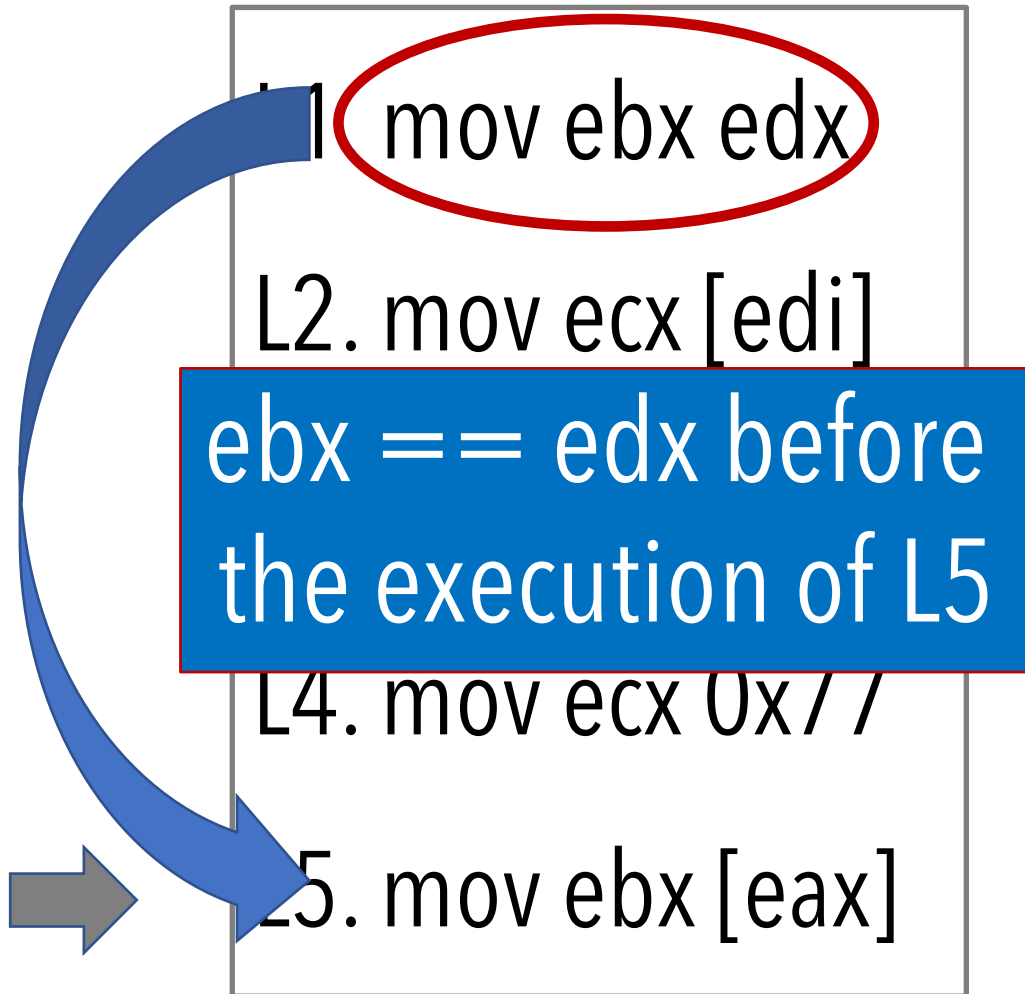
Postmortem Program Analysis

- Perform data flow analysis against the identified path.
 - Reconstruct the data flow that a program followed prior to its crash.
 - Examine how a bad value was passed to the crash site.
- Key problem in reconstructing data flow: alias detection.
 - Alias: two pointers point to the same memory location.



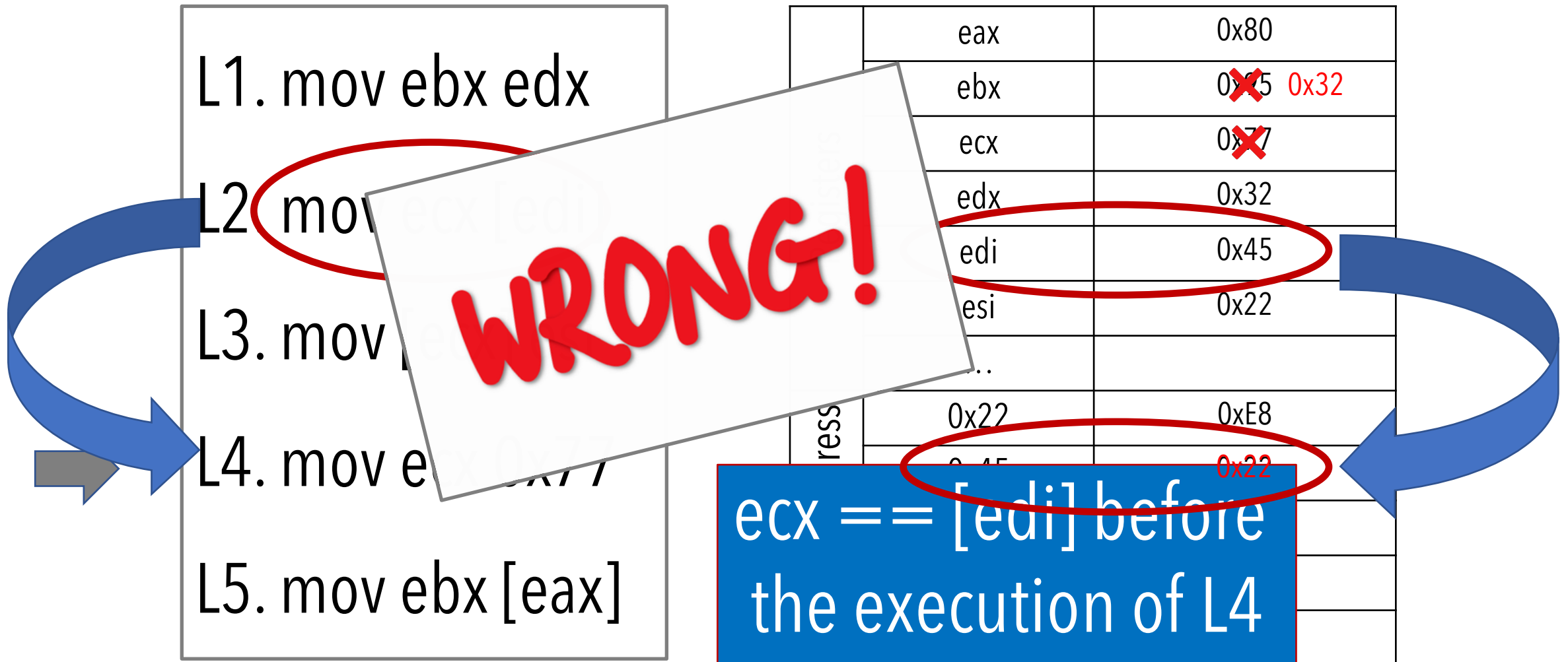
Data flow

Reverse Execution with Backward Data Flow Analysis



Registers	eax	0x80
	ebx	0x95
	ecx	0x77
	edx	0x32
	edi	0x45
	esi	0x22
	...	
Memory Address	0x22	0xE8
	0x45	0x22
	0x77	0xB7
	0x80	0x95
	...	

Challenge – Memory Alias Issue



Challenge – Memory Alias Issue

Pointing
to the same
memory
region

L1. mov ebx edx

L2. mov ecx [edi]

L3. mov [ecx] esi

L4. mov ecx 0x77

L5. mov ebx [eax]

Registers	eax	0x80
	ebx	0x45 0x32
	ecx	0x77
	edx	0x32
	edi	0x45
	esi	0x22
	...	
ress	0x22	0xE8
	0x45	0x32

edi == ecx right before
the execution of L4



Alias Analysis

- Current postmortem program analysis system.
 - Adopt hypothesis testing to detect alias.
 - Extremely slow: spend weeks to find the root cause for a program.
 - Even slower than human analyst.
- Goal: accurate and efficient alias analysis.

Roadmap

- Value-set Analysis.
- Challenges of VSA in postmortem program analysis.
- Design overview.
- Technical details.
- Evaluation in real world crashes.
- Summary.



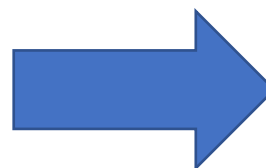
Value-set Analysis

- State-of-the-art binary level alias analysis technique.
- Being integrated into a variety of binary analysis frameworks (Angr, BAP).
- High level idea:
 - Given a control flow, VSA first assigns each instruction into different memory region (Heap, Stack, global).
 - Tracks down *a-locs*: register, memory call on stack, heap or global.
 - Compute a value set for each a-loc: (global, stack, heap).
 - Identify memory alias according to the value sets.

Value-set Analysis

- Demonstration of Value-set Analysis.

1	sub	esp, 44	Truncated Trace
2	lea	eax, [esp+4]	
3	lea	ebx, [esp+24]	
4	mov	[esp+0], eax	
5	mov	ecx, 0	
<hr/>			
L1	mov	edx, [0xC4]	
7	mov	[eax], edx	
8	mov	edx, [0xC8]	
9	mov	[ebx], edx	
10	add	eax, 4	
11	add	ebx, 4	
12	inc	ecx	
13	cmp	ecx, 1	
14	j1	L1	
15	mov	[esp+4], ecx	



Complete Trace		
	<i>A-loc</i>	Value-set
1	esp	(⊥, [-44, -44], ⊥)
2	eax	(⊥, [-40, -40], ⊥)
3	ebx	(⊥, [-20, -20], ⊥)
4	[esp] (⊥, [-44, -44], ⊥)	(⊥, [-40, -40], ⊥)
5	ecx	([0, 0], ⊥, ⊥)
L1	[0xC4] ([0xC4, 0xC4], ⊥, ⊥)	([0, 0], ⊥, ⊥)
	edx	([0, 0], ⊥, ⊥)
7	[eax] (⊥, [-40, -40], ⊥)	([0, 0], ⊥, ⊥)
8	[0xC8] ([0xC8, 0xC8], ⊥, ⊥)	([0, 0], ⊥, ⊥)
	edx	([0, 0], ⊥, ⊥)
9	[ebx] (⊥, [-20, -20], ⊥)	([0, 0], ⊥, ⊥)
10	eax	(⊥, [-36, -36], ⊥)
11	ebx	(⊥, [-16, -16], ⊥)
12	ecx	([1, 1], ⊥, ⊥)
13	-	-
14	-	-
15	[esp+4] (⊥, [-40, -40], ⊥)	([1, 1], ⊥, ⊥)

Why not Value-set Analysis in Postmortem Analysis

- In the content of postmortem program analysis:
 - The full control flow or execution trace is not available.
 - Core dump can only record limit length of execution trace.
 - VSA will perform over-approximation in value-set construction.

```
L1      mov     edx,[0xC4]
7       mov     [eax],edx
8       mov     edx,[0xC8]
9       mov     [ebx],edx
10      add     eax,4
11      add     ebx,4
12      inc     ecx
13      cmp     ecx,1
14      jl      L1
15      mov     [esp+4],ecx
```



Incomplete Trace	
<i>A-loc</i>	Value-set
NA	NA
NA	NA
NA	NA
NA	NA
NA	NA
[0xC4] ([0xC4, 0xC4], ⊥, ⊥)	(T, T, T)
edx	(T, T, T)
[eax] (T, T, T)	(T, T, T)
[0xC8] ([0xC8, 0xC8], ⊥, ⊥)	(T, T, T)
edx	(T, T, T)
[ebx] (T, T, T)	(T, T, T)
eax	(T, T, T)
ebx	(T, T, T)
ecx	(T, T, T)
-	-
-	-
[esp+4] (⊥, [4, 4], ⊥)	(T, T, T)

Why not Value-set Analysis in Postmortem Analysis

- Alias Analysis Results.
 - Complete trace: **100%** correctly identify the alias pairs.
 - Incomplete trace: mark **60%** of the memory pairs as may-alias.
 - Over-approximation.

Incomplete trace

	[esp]	[0xC4]	[0xC8]	[eax]	[ebx]	[esp+4]
[esp]	-	0	0	0	0	0
[0xC4]	NA	-	0	0	0	0
[0xC8]	NA	0	-	0	0	0
[eax]	NA	?	?	-	0	1
[ebx]	NA	?	?	?	-	0
[esp+4]	NA	?	?	?	?	-



Why Deep Learning

- Previous applications demonstrated that DL can be used to learn patterns from input and assign each input an label.
- For alias analysis:
 - Input data (instructions): a sequence of machine code.
 - Labels: memory regions each instruction is attached.
 - E.g. `push 0x68732f2f`: data [68 2f 2f 73 68].
Label [stack stack stack stack stack].

Why Deep Learning

- Learn from the previous sequence that reflects the process of determining memory region access.
- Memory region that an instruction accesses can be determined:
 - Semantics of that instruction.
 - `push eax`: indicate a stack access.
 - Context indicated by previous instruction.

```
0: 8d 1c 24 lea ebx,[esp]
```

```
3: 89 0b mov DWORD PTR [ebx],ecx
```

→ [ebx] indicates a stack access.

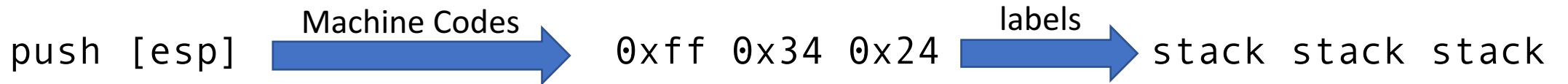


Challenges for DL

- Capture the sequential dependence within input sequence.
 - Each hex in the binary code sequence are highly correlated with each other.
- The bi-directional dependence.
 - Forward analysis procedure.
 - Backward analysis procedure.

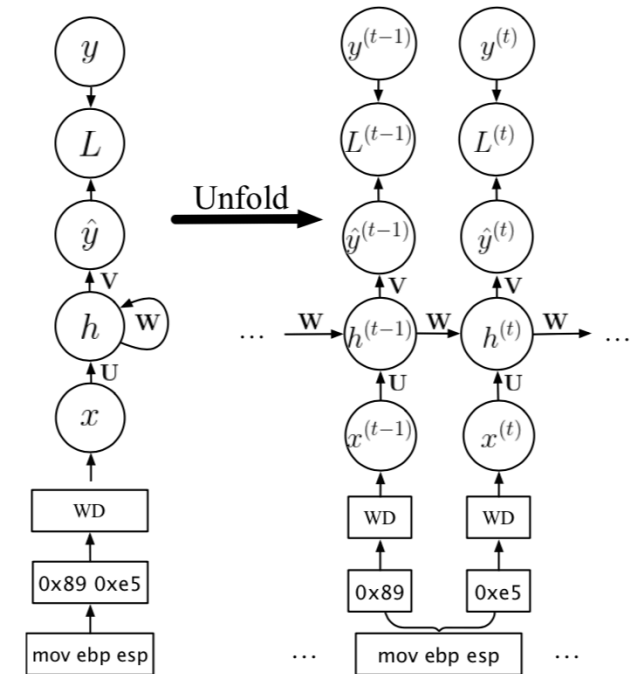
Challenges for DL

- Catch the dependence between not only the dependency between and within instructions (input sequence) but also dependencies between adjacent labels.
 - Hexes in the same instruction have the same label.



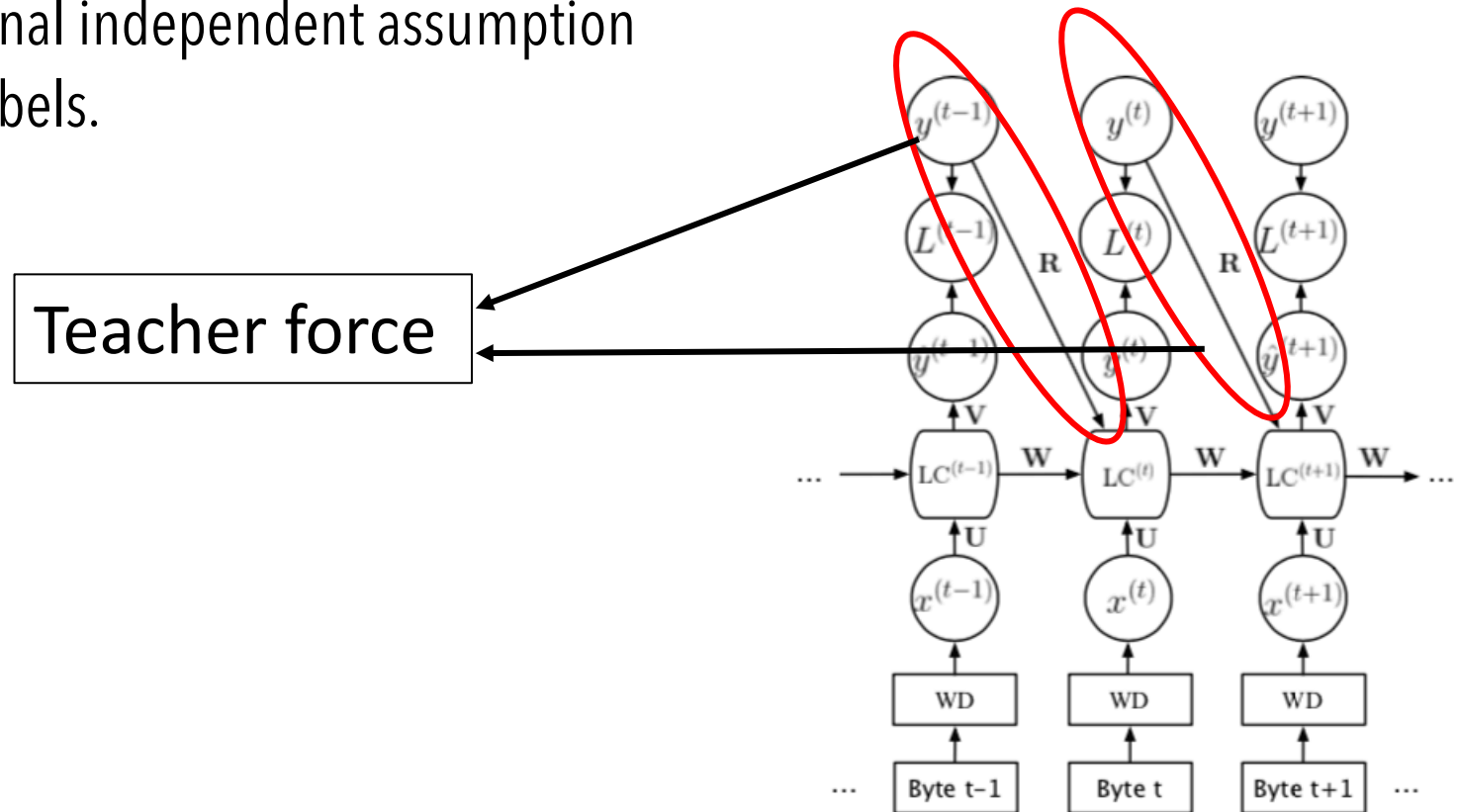
Recurrent neural network – dependence within Hexes

- Recurrent neural network.
 - Take sequential data as input.
 - self-connected hidden units.
 - Model the dynamic temporal behavior for a time sequence.
 - Types of hidden units:
 - SimpleRNN.
 - GRU.
 - LSTM (Capture long term dependence).



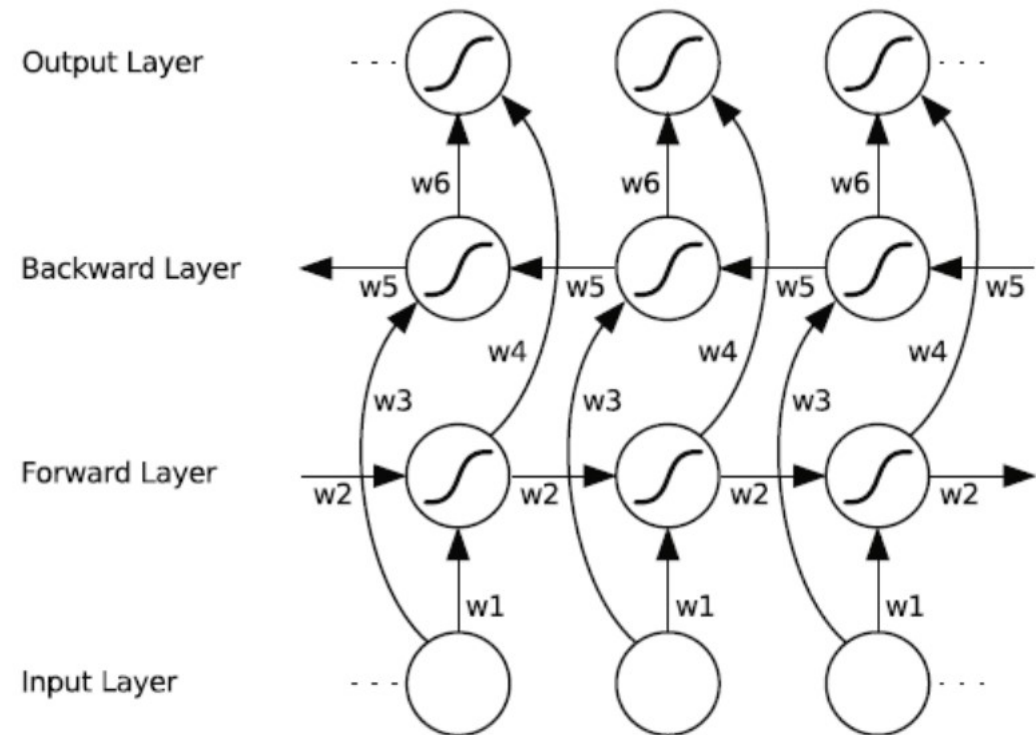
Teacher force – dependence between labels

- Integrate the label of late time step into the current input.
 - Remove the conditional independent assumption between adjacent labels.



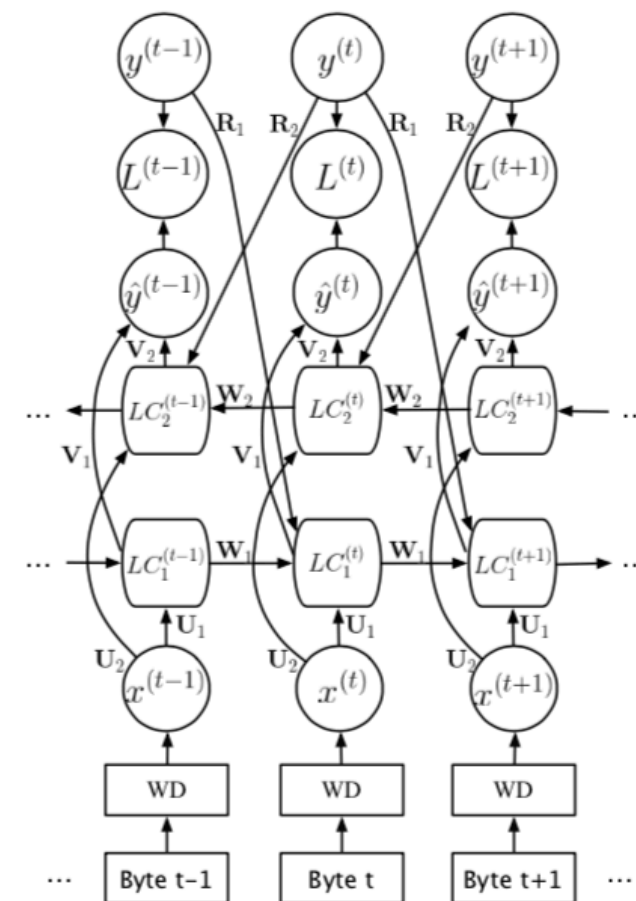
Bi-directional RNN – backward dependence

- Bi-directional RNN combines a RNN that moves forward, beginning from the start of the byte sequence, with another RNN that moves backward.



DEEPVSA: Deep learning Facilitated VSA

- Novel network structure: Bi-directional conditional LSTM.
 - Integrate LSTM, teacher force and Bi-directional connection together.
- Extension of conventional VSA.
 - Take the region predictions of DL model to determine non-alias relationships that the conventional VSA originally fails to identify.





Evaluation

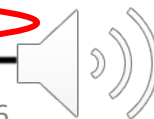
- Experimental setup.
 - Dataset:
 - Randomly select 54 vulnerability reports from the Exploit Database Archive.
 - Run the PoC programs tied to corresponding vulnerabilities, trigger software failures and collect the 54 execution traces pertaining to the crashes.
 - Answer the following questions:
 - Does the bi-directional conditional LSTM exhibit better performance than other RNN architecture?
 - Can the memory regions identified improve the memory alias analysis detection performance?

Question one

- Comparison between different RNN structures.
 - Label the bytes tied to the execution traces with the memory regions.
 - Randomly divided the traces into 5 disjoint groups and train 5 distinct models for each network architectures.
 - Each model: take one group of traces as testing data and utilize the remaining to train our neural network models.

- Our model established highest precision and recall.

		Global	Heap	Stack	Other
Precision	Bi-RNN	99.55%	99.33%	99.96%	99.75%
	Bi-GRU	99.55%	99.49%	99.98%	99.80%
	Bi-LSTM	99.55%	99.30%	99.97%	99.76%
	Our model	99.99%	99.79%	99.99%	99.88%
Recall	Bi-RNN	99.50%	99.47%	99.94%	99.81%
	Bi-GRU	99.54%	99.49%	99.94%	99.78%
	Bi-LSTM	99.51%	99.55%	99.94%	99.81%
	Our model	99.88%	99.76%	99.97%	99.90%



Question two

- Memory alias pairs detection.
 - Apply the DL model to predict the memory regions tied to each trace.
 - Pass the predictions to DEEPVSA and compute the non-alias pairs.
- Use conventional VSA as baseline for comparison.

Program	CVE/EDB-ID	LoC	TR?	VSA		DEEPVSA		Statistics			
				Non-alias	Error rate	Non-alias	Error rate	Global	Heap	Stack	Other
DXFScope-0.2	CVE-2004-1271	7697	✗	47.33%	0%	86.06%	0%	5	0	647	338
autotrace-0.31.1	CVE-2017-9180	12620	✗	54.90%	0%	93.30%	0%	75	517	12804	4801
bento4-1.5.0-617	CVE-2017-14638	43610	✗	58.73%	0%	96.14%	0%	254	5335	28428	4986
psutils-p17	EDB-890	1736	✗	29.96%	0%	81.73%	0%	47	88	21492	16680
gif2png-2.5.2	CVE-2009-5018	1331	✗	25.34%	0%	78.18%	0%	17	770	24040	16445
bento4-1.5.0-617	CVE-2017-14640	43610	✗	56.53%	0%	95.18%	0%	351	7784	37953	7380
openjpeg-2.1.1	CVE-2016-7445	169538	✗	36.38%	0%	84.93%	0%	159	195	28214	19736
libpng-1.2.5	CVE-2004-0597	33681	✗	37.37%	0%	85.88%	0%	30	1109	30908	21366
unrtf-0.19.3	CVE-2004-1297	5039	✗	35.84%	0%	87.52%	0%	994	5905	32803	23018
LibSMI-0.4.8	CVE-2010-2891	80461	✗	51.10%	0%	91.84%	0%	8	2638	50213	21606
libzip-1.2.0	CVE-2017-12858	37083	✗	17.91%	0%	80.71%	0%	17	15966	35285	21126
TestDisk-6.14	EDB-36881	64345	✓	34.62%	0%	83.47%	0%	262	29644	44752	52369
unalz-0.52	CVE-2005-3862	8546	✓	4.00%	0%	55.46%	0%	69	6229	52353	28511
JPegToAvi-1.5	CVE-2004-1279	580	✓	0.01%	0%	49.96%	0%	28	336	52363	46879
o3read-0.0.3	CVE-2004-1288	932	✓	28.00%	0%	79.00%	0%	3031	0	61488	63950
corehttp-0.5.3.1	CVE-2009-3586	935	✓	63.44%	0%	95.61%	0%	564	9905	92576	27357
corehttp-0.5.3alpha	CVE-2007-4060	935	✓	63.90%	0%	95.67%	0%	564	10977	96820	28616
mp3info-0.8.5a	CVE-2006-2465	3212	✓	0.03%	0%	50.74%	0%	138	4564	86747	58607
bento4-1.5.0-617	CVE-2017-14641	43610	✓	37.26%	0%	84.31%	0%	507	36751	85352	19009
unrar-3.9.3	EDB-17611	17575	✓	69.72%	0%	90.46%	0%	5793	0	90038	30720
HTML2HDML-1.0.3	CVE-2004-1275	7894	✓	28.29%	0%	80.19%	0%	192	28120	35438	4200
SQLite-3.8.6	CVE-2015-5895	98039	✓	0.13%	0%	43.72%	0%	196	9571	67889	22893
htmldoc-1.8.27	CVE-2009-3050	59237	✓	0.15%	0%	45.29%	0%	382	5256	71160	24923
sudo-1.8.0	CVE-2012-0809	38761	✓	0.11%	0%	50.29%	0%	81	8892	64057	27490
GnuPG-1.9.14	CVE-2006-3746	99053	✓	32.16%	0%	83.00%	0%	1	45073	44988	2784
gas-2.12	CVE-2005-4807	595504	✓	3.41%	0%	65.09%	0%	2369	33388	31986	10309
mcrypt-2.5.8	CVE-2012-4409	37439	✓	0.03%	0%	37.82%	0%	49	1955	49921	16862
nasm-0.98.38	CVE-2004-1287	33553	✓	48.68%	0%	92.96%	0%	5441	27841	56075	9899
prozilla-1.3.6	CVE-2004-1120	9000	✓	0.01%	0%	54.00%	0%	149	25455	50780	11355
stftp-1.1.0	EDB-9264	1559	✓	43.70%	0%	90.14%	0%	1647	32650	52504	7850
gdb-7.5.1	EDB-23523	1665735	✓	2.72%	0%	60.61%	0%	1268	46335	49770	10058
Overkill-0.16	CVE-2006-2971	16361	✓	5.56%	0%	11.00%	0%	0	0	4652	74418
gdb-6.6	EDB-30142	1377746	✓	15.87%	0%	62.21%	0%	3654	13147	69093	22559
make-3.81	EDB-34164	24168	✓	0.01%	0%	19.93%	0%	57	25700	2886	867
coreutils-8.4	CVE-2013-0223	138135	✓	81.50%	0%	96.00%	0%	1524	4578	110193	8377
ClamAV-0.93.3	CVE-2008-5314	69430	✓	99.99%	0%	100.00%	0%	0	0	118264	16179
ClamAV-0.88.2	CVE-2006-5295	39052	✓	38.53%	0%	90.00%	0%	0	32758	59916	27333
ClamAV-0.88.2	CVE-2006-4018	39052	✓	30.60%	0%	78.03%	0%	0	43830	0	64745
putty-0.66	CVE-2016-2563	90165	✓	32.52%	0%	71.52%	0%	2704	17304	63094	10987
coreutils-8.4	CVE-2013-0222	138135	✓	0.39%	0%	19.03%	0%	30	74	5273	73534
autotrace-0.31.1	CVE-2017-9182	12620	✓	45.24%	0%	62.00%	0%	435	3045	112323	10878

Question two

- Memory alias pairs detection.
 - DEEPVSA improves the detection rate from 24.84% to 66.43%.
 - DEEPVSA dose not introduce errors.

Program	CVE/EDB-ID	LoC	TR?	VSA		DEEPVSA		Statistics			
				Non-alias	Error rate	Non-alias	Error rate	Global	Heap	Stack	Other
GnuPG-1.9.14	CVE-2006-3082	99053	✓	32.20%	0%	83.52%	0%	0	45069	45072	2818
podofo-0.94	CVE-2017-5854	60147	✓	0.49%	0%	27.86%	0%	1104	15285	105954	6200
Average	-	-	-	24.84%	0%	66.43%	0%	728	14575	49670	32795



Summary

- DEEPVSA implements a novel RNN architecture customized for VSA.
 - Bi-directional conditional LSTM.
- DEEPVSA outperforms the off-the-shelf recurrent network architecture in terms of memory region identification.
- DEEPVSA significantly improves the VSA with respect to its capability in analyzing memory alias.
- DEEPVSA will enhance the **accuracy** and **efficiency** of the postmortem program analysis .

Thank you very much!

