

Fine Grained Dataflow Tracking with Proximal Gradients

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Dataflow Analysis

Summarize dataflows for **all possible inputs**

```
//input x
```

```
int x1 = 2 * x;
```

```
int x2 = x1 % 4;
```

```
return x2;
```

Dataflow Summary:

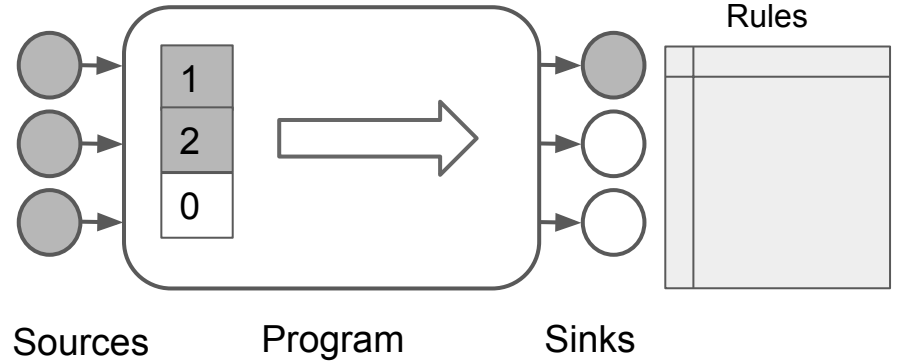
even inputs \Rightarrow return 0

odd inputs \Rightarrow return 2

Dynamic Taint Analysis (DTA)

Uses boolean **Taint Labels**

Applies **Propagation Rules** during Execution



Used in many security applications:



Exploit Detection



Greybox Fuzzing



Malware Analysis



Information Leak
Identification

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```
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int x1 = 2 * x;
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return x2;
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Dynamic Taint Analysis (DTA)

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```
//input x  
  
int x1 = 2 * x;  
int x2 = x1 % 4;  
  
return x2;
```

Label 1



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```
//input x
```

```
int x1 = 2 * x;
```

```
int x2 = x1 % 4;
```

```
return x2;
```

Apply Rule

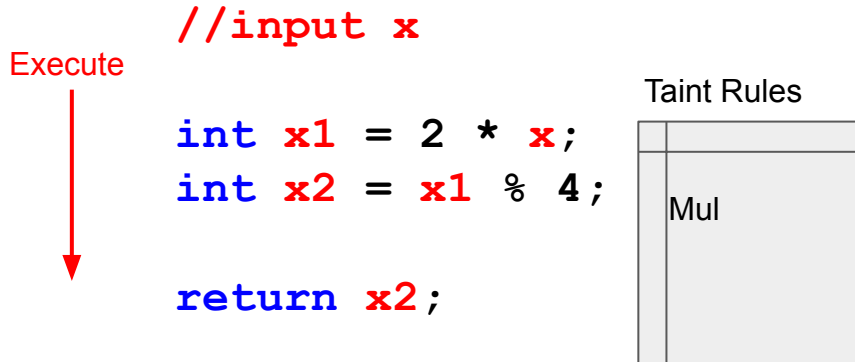
Taint Rules



Dynamic Taint Analysis (DTA)

Uses boolean **Taint Labels**

Applies **Propagation Rules** during Execution



Dataflow Summary:

input x \Rightarrow can change x2

Dynamic Taint Analysis (DTA)

Boolean Taint labels **cannot represent** all possible behavior

Causes taint **propagation errors**

Dynamic Taint Analysis (DTA)

Boolean Taint labels **cannot represent** all possible behavior

Causes taint **propagation errors**

```
//input x, k  
  
int x1 = x - k;  
int x2 = x - x1;  
  
return x2;
```

Dynamic Taint Analysis (DTA)

Boolean Taint labels **cannot represent** all possible behavior

Causes taint **propagation errors**

```
//input x, k
```

```
int x1 = x - k;
```

```
int x2 = x - x1; ← x - x + k
```

```
return x2;
```

Dynamic Taint Analysis (DTA)

Boolean Taint labels **cannot represent** all possible behavior

Causes taint **propagation errors**

```
//input x, k
```

```
int x1 = x - k;
```

```
int x2 = x - x1;
```

```
return x2; ← taint error!
```

Dynamic Taint Analysis (DTA)

Boolean Taint labels **cannot represent** all possible behavior

Causes taint **propagation errors**

Sound Rules



OverTaint Errors, False Alerts

Precise Rules



UnderTaint Errors, Missed Violations

Chua, Zheng Leong, et al. "One Engine To Serve'em All: Inferring Taint Rules Without Architectural Semantics." *NDSS*. 2019.

Dalton, Michael, Hari Kannan, and Christos Kozyrakis. "Tainting is not pointless." *ACM SIGOPS Operating Systems Review* 44.2 (2010): 88-92.

Slowinska, Asia, and Herbert Bos. "Pointer tainting still pointless: (but we all see the point of tainting)." *ACM SIGOPS Operating Systems Review* 44.3 (2010): 88-92.

Yadegari, Babak, and Saumya Debray. "Bit-level taint analysis." *2014 IEEE 14th International Working Conference on Source Code Analysis and Manipulation*. IEEE, 2014.

Richer form of Dataflow?

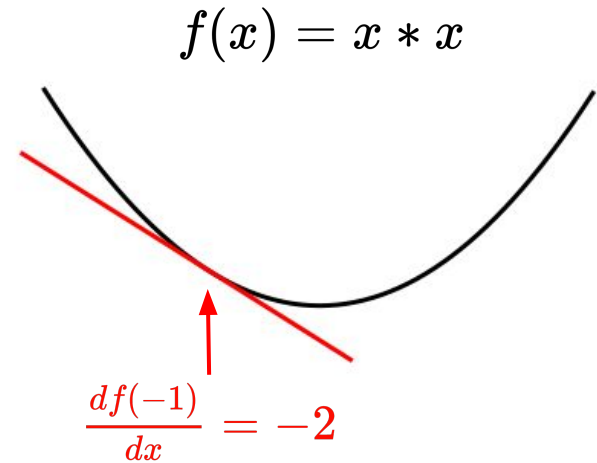
Richer form of Dataflow?

Gradient is a popular **Dataflow** measure in machine learning:

Richer form of Dataflow?

Gradient is a popular **Dataflow** measure in machine learning:

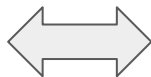
$$\frac{df(x)}{dx} = \frac{\text{change in output}}{\text{change in input}}$$



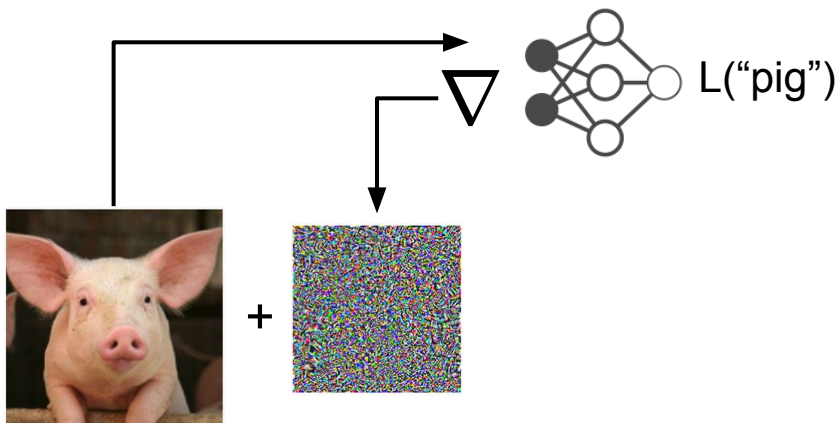
Richer form of Dataflow?

Gradient used for similar applications to Dataflow:

- Guiding adversarial testing



Vulnerability search:



```
int y = x1 + x2%2;  
  
if (y > THRESHOLD) {  
    // vulnerability  
}
```


Gradient as Dataflow Measure

Gradient can be composed with **chain rule**

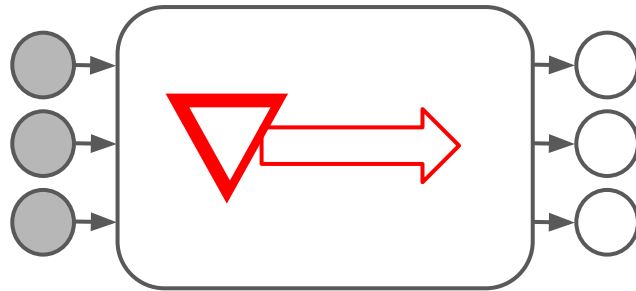
$$\frac{df(g(x))}{dx} = \frac{df}{dg} * \frac{dg}{dx}$$

Gradient as Dataflow Measure

Gradient can be composed with **chain rule**

$$\frac{df(g(x))}{dx} = \frac{df}{dg} * \frac{dg}{dx}$$

Key Idea: Propagate Gradient directly over Program



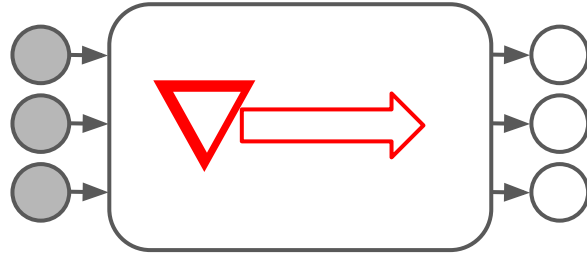
Sources

Program

Sinks

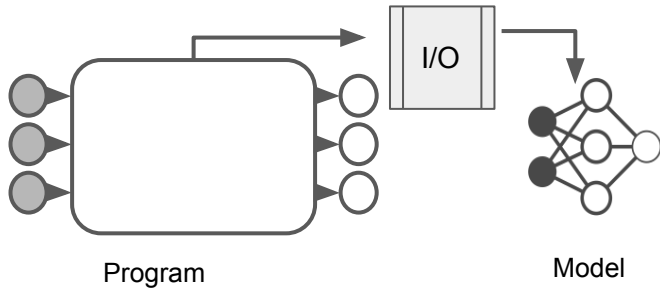
A gray rectangular box with a thin border, containing the mathematical expression $\frac{df}{dg} * \frac{dg}{dx}$ in red text.

Gradient as Dataflow Measure

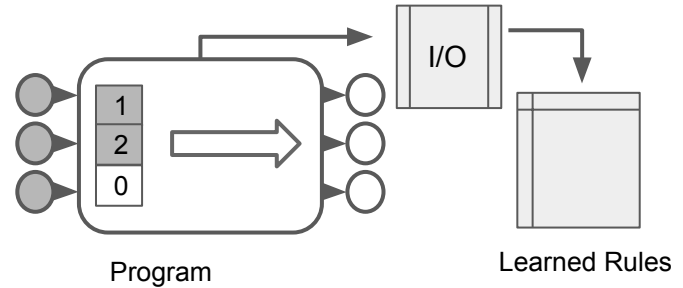


$$\frac{df}{dg} * \frac{dg}{dx}$$

Neutaint: Measures Gradient of Neural Network



TaintInduce: Learns taint rules from I/O samples



Gradient Reduces Errors

Propagates dataflow correctly

```
//input x, k
```

```
int x1 = x - k;
```

```
int x2 = x - x1; ← x - x + k
```

```
return x2;
```

Gradient Reduces Errors

Propagates dataflow correctly

```
//input x, k
```

```
int x1 = x - k;  $\frac{dx1}{dx} = 1$ 
```

```
int x2 = x - x1;  $\frac{dx2}{dx} = 1 - \frac{dx1}{dx} = 0$ 
```

```
return x2;
```

Gradient Reduces Errors

Propagates dataflow correctly

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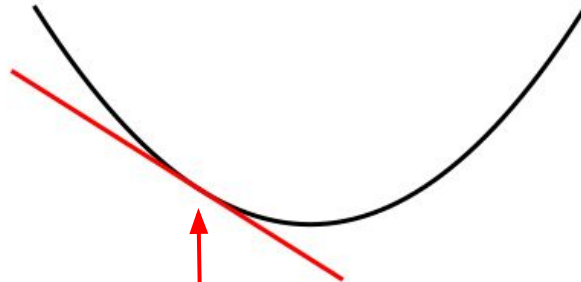
```
int x1 = x - k;  $\frac{dx1}{dx} = 1$ 
```

```
int x2 = x - x1;  $\frac{dx2}{dx} = 1 - \frac{dx1}{dx} = 0$ 
```

```
return x2; ← gradient to x is 0!
```

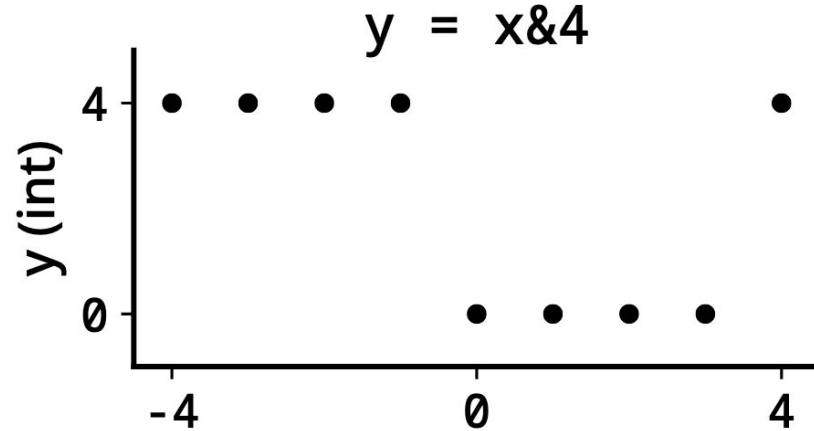
Problem: Programs are not differentiable!

$$f(x) = x * x$$

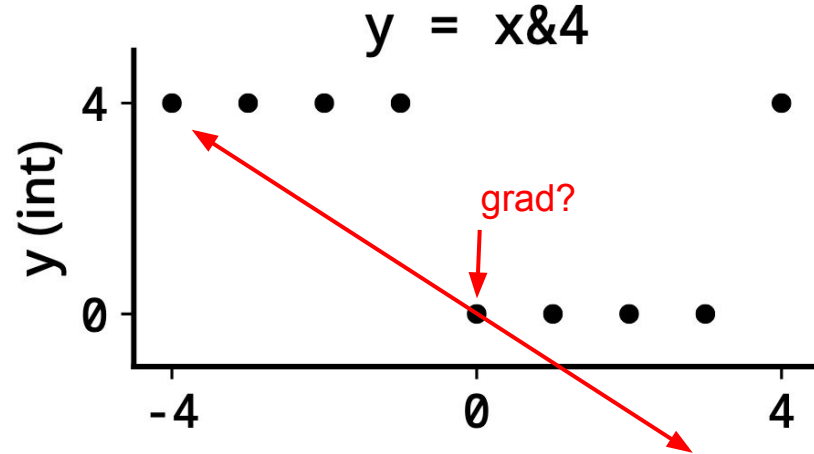


$$\frac{df(-1)}{dx} = -2$$

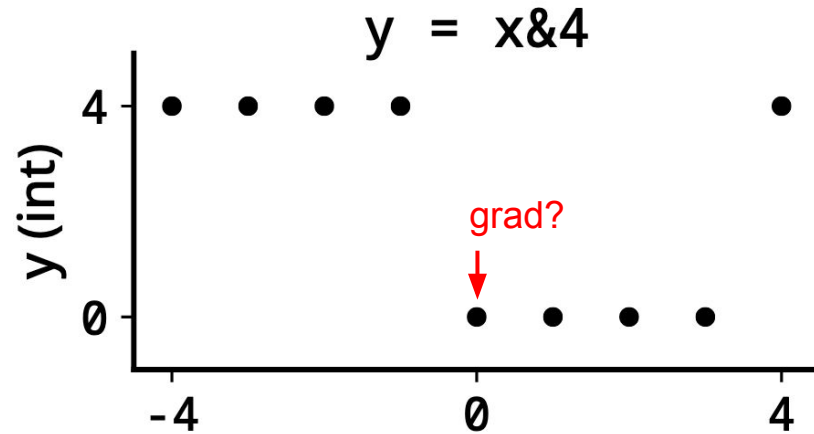
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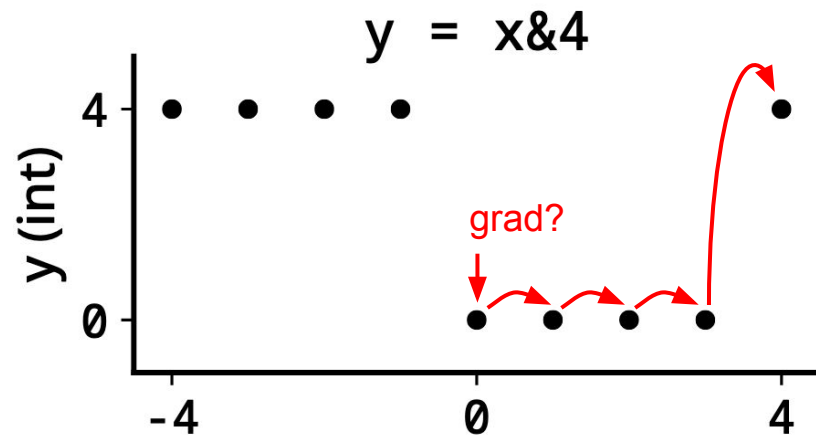
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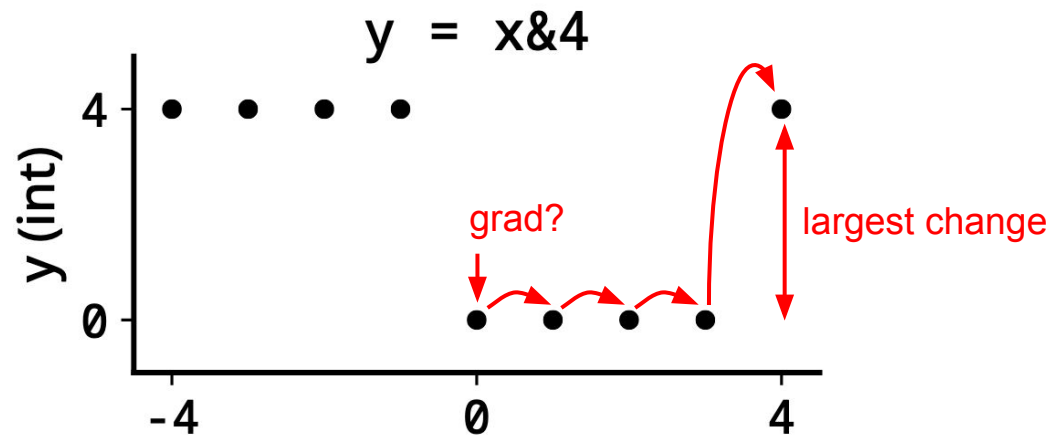
Proximal Gradient



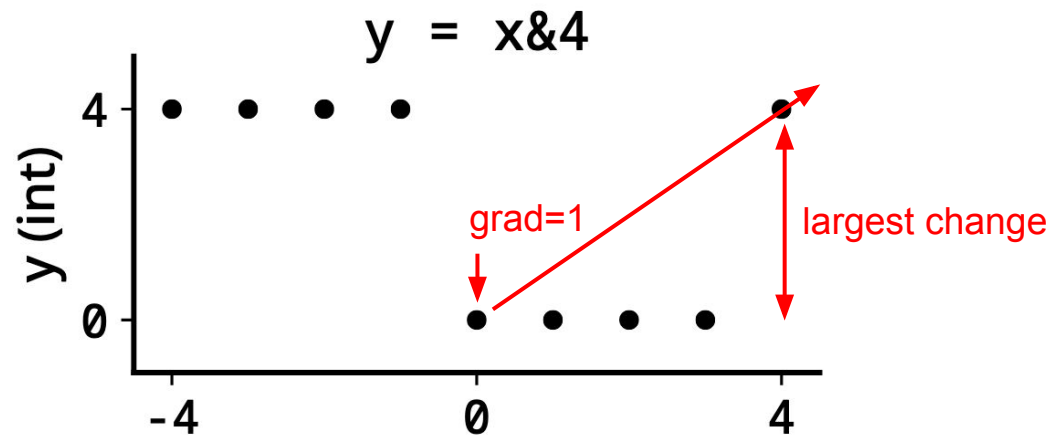
Proximal Gradient



Proximal Gradient



Proximal Gradient



Proximal Gradient Analysis (PGA)

Implement as **LLVM Sanitizer Pass** (`grsan`)

Evaluate **Dataflow Accuracy**:

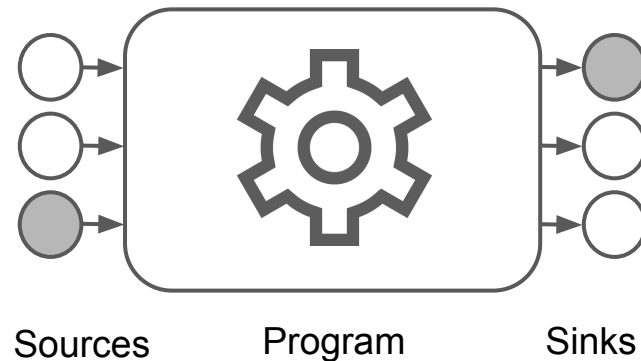
Ground Truth Dataflow:

For each input byte:

- flip each bit

- set to 0 and 255

- record changed sink variables



Proximal Gradient Analysis (PGA)

Implement as **LLVM Sanitizer Pass (grsan)**

Evaluate **Dataflow Accuracy**:

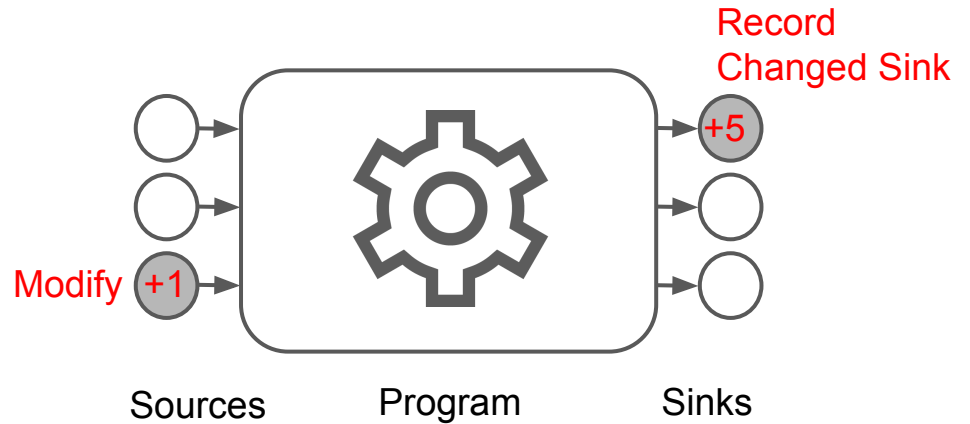
Ground Truth Dataflow:

For each input byte:

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Proximal Gradient Analysis (PGA)

	libdft			dfsan			grsan (floats)		
	Prec.	Rec.	F1	Prec.	Rec.	F1	Prec.	Rec.	F1
minigzip	0.42	0.29	0.17	0.29	0.60	0.39	0.63	0.51	0.57
djpeg	-	-	-	0.22	1.00	0.37	0.60	0.83	0.69
mutool	0.70	0.32	0.22	0.63	0.61	0.62	0.86	0.51	0.64
xmllint	-	-	-	0.62	0.99	0.76	0.94	0.91	0.92
objdump	0.47	0.67	0.28	0.37	0.93	0.52	0.66	0.77	0.71
strip	0.26	0.59	0.18	0.20	0.96	0.33	0.50	0.86	0.63
size	0.20	0.59	0.30	0.37	0.95	0.53	0.62	0.91	0.74

F1 accuracy improvement: Up to **33% over SOTA** (20% better on average)

Proximal Gradient Analysis

Introduce **proximal gradients (PGA)** as new formulation of dataflow problem

Show PGA **improves dataflow accuracy**, and gradients are useful in dataflow applications.

To learn more about PGA please see our paper:

Paper: <https://www.usenix.org/system/files/sec21fall-ryan.pdf>

Code: <https://github.com/gryan11/PGA>