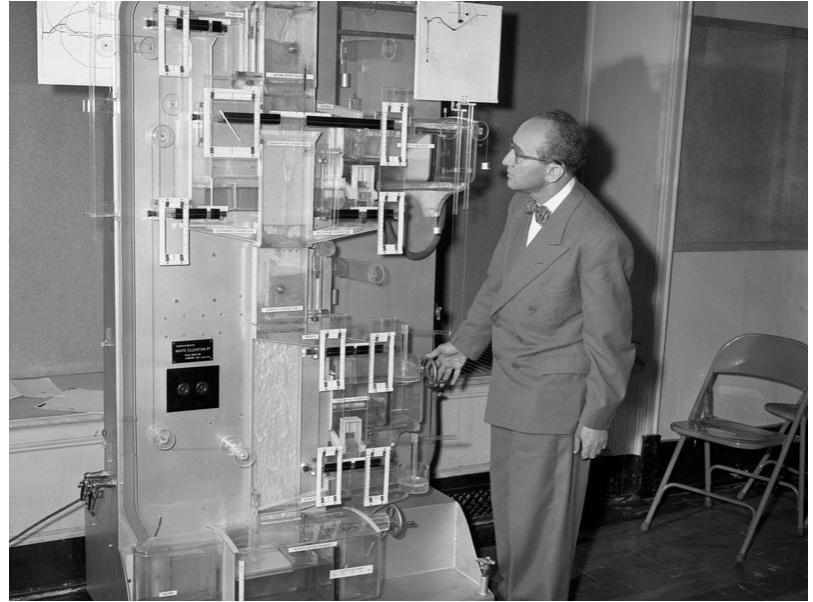
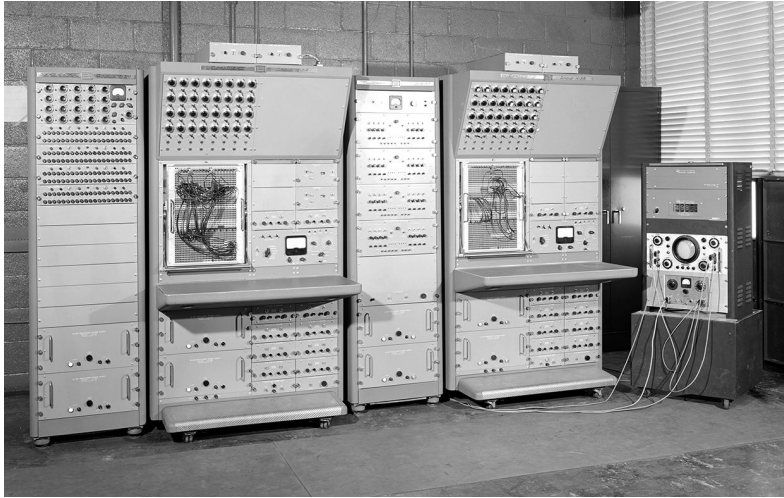




New Compilation Techniques for Reconfigurable Analog Devices

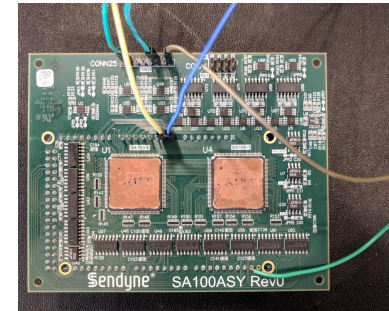
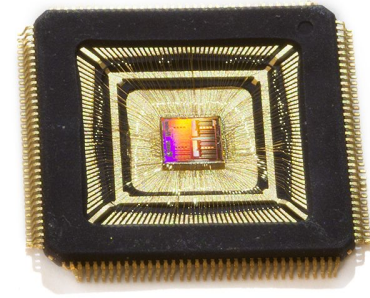
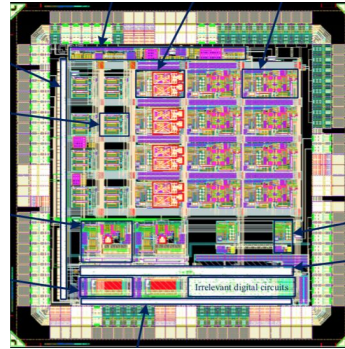
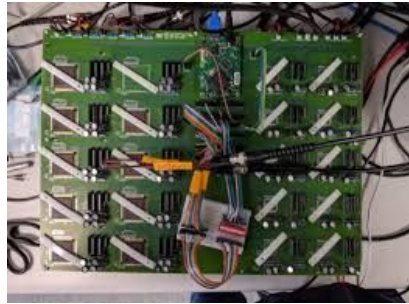
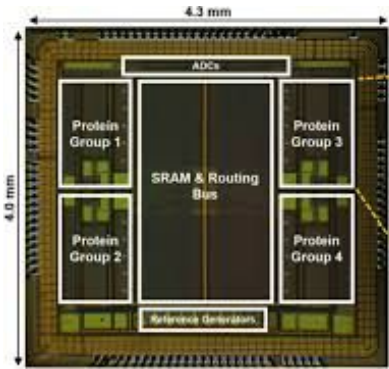
Sara Achour
Stanford CS & Stanford EE

**Automatically program analog
computers to perform computation**

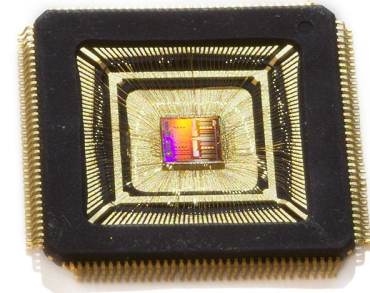
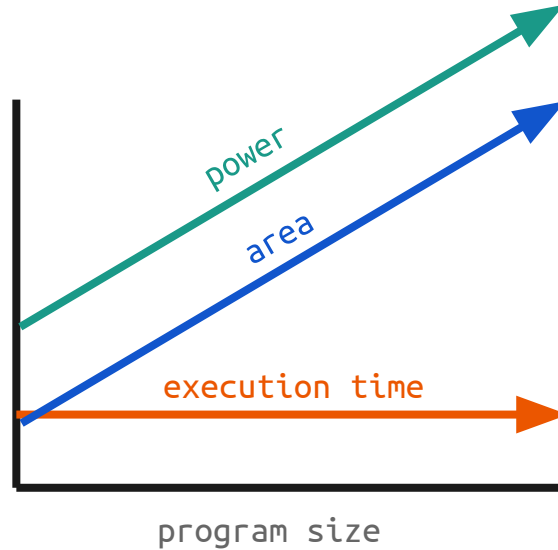


Ultra-Low Power Reconfigurable Analog Devices

- CMOS Technology
- Digitally reconfigurable

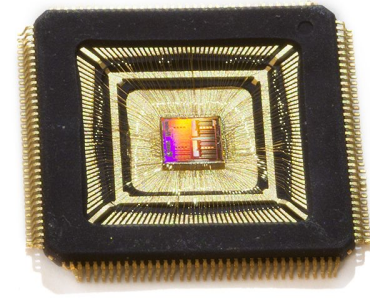
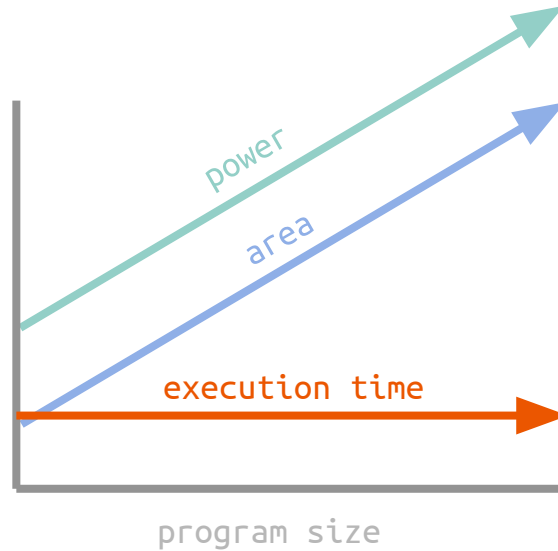


Ultra-Low Power Reconfigurable Analog Devices



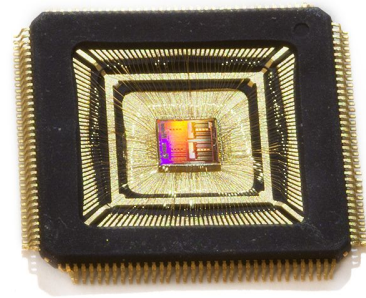
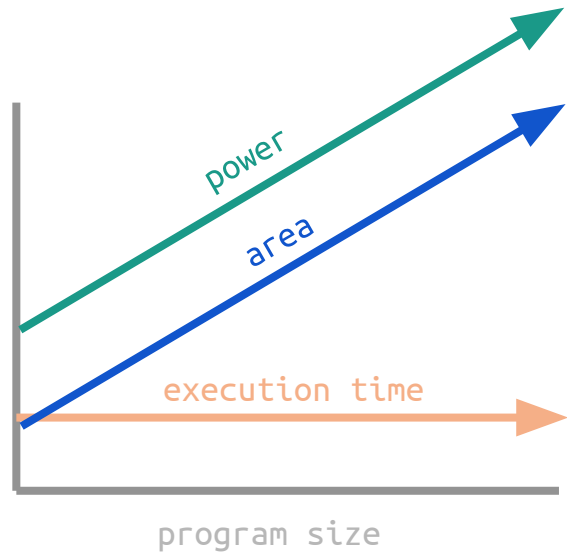
Perform computation using microjoules of energy!

Ultra-Low Power Reconfigurable Analog Devices



Execution time is constant w.r.t. complexity of computation

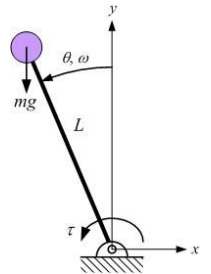
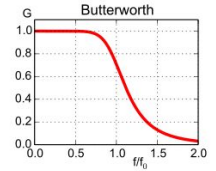
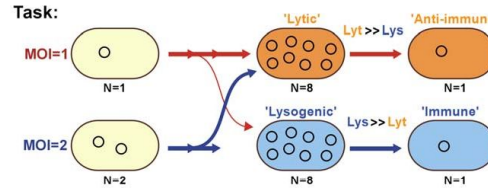
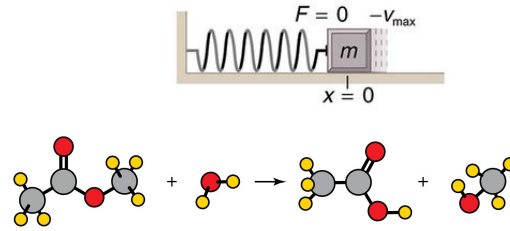
Ultra-Low Power Reconfigurable Analog Devices



**This target class of analog
devices executes dynamical
systems**

Differential-Equation Solving Analog Devices

What are Dynamical Systems?

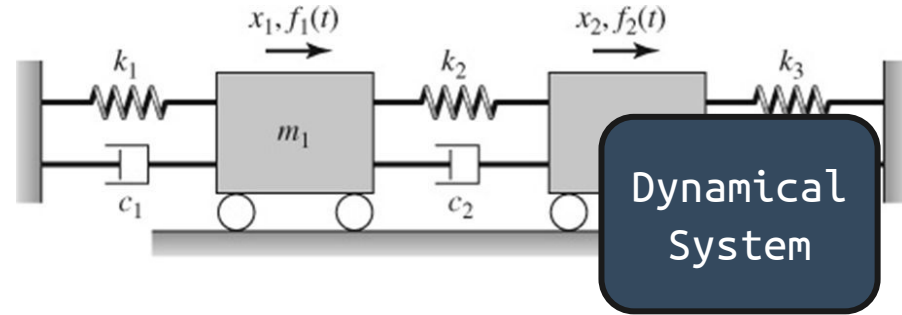


Differential-Equation Solving
Analog Devices

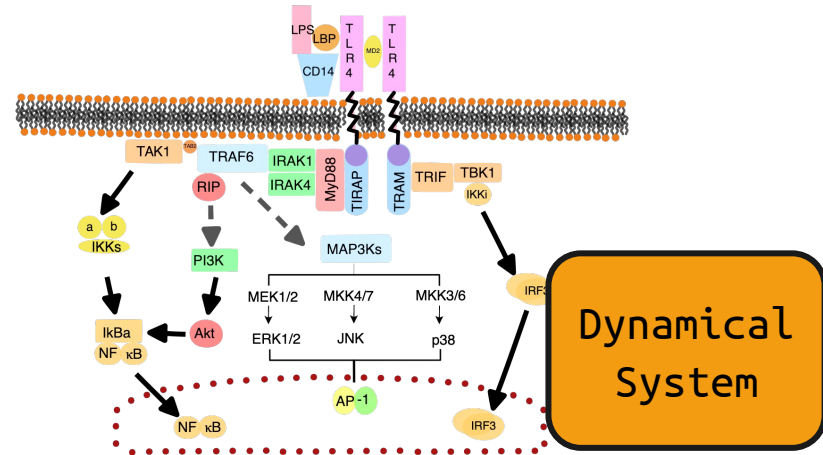
Dynamical Systems
can be used to

**Study physical
phenomena**

Two-Spring System



Biological Pathway

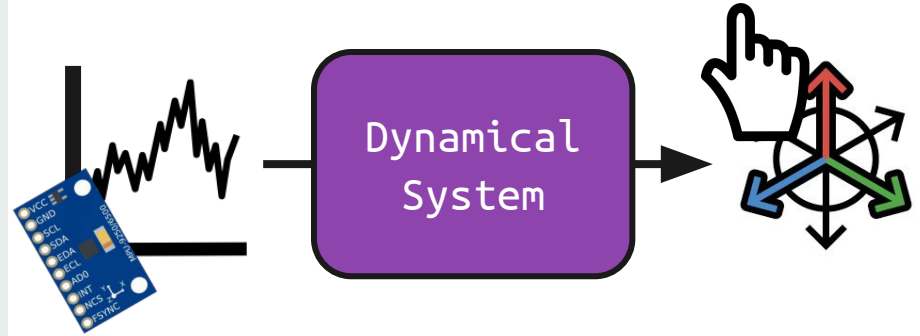


Differential-Equation Solving
Analog Devices

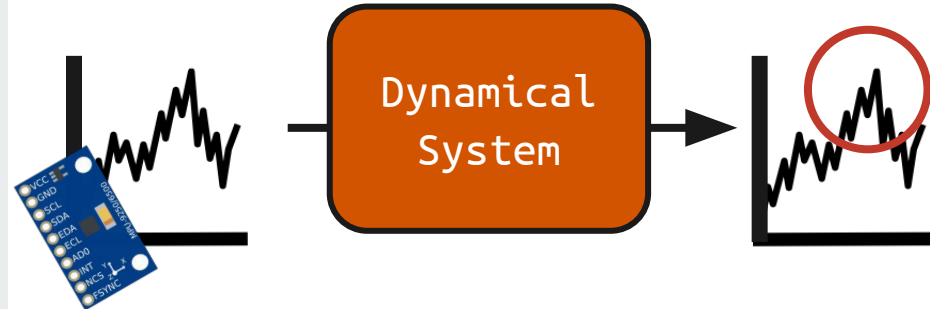
**Dynamical Systems
can be used to**

**Analyze
surroundings**

Position Tracking



Anomaly Detection

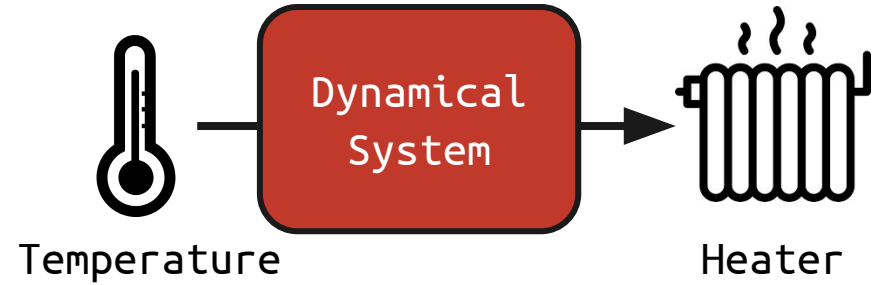


Differential-Equation Solving
Analog Devices

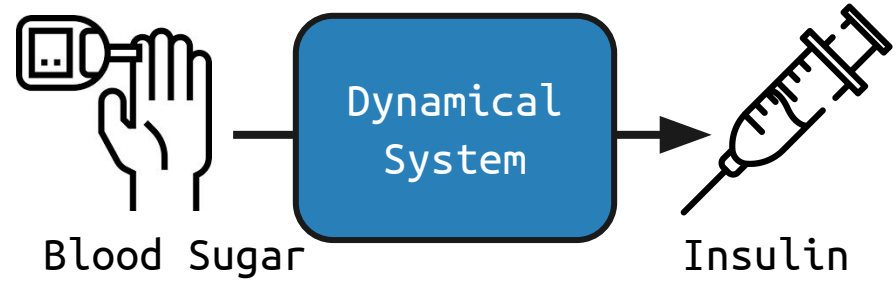
**Dynamical Systems
can be used to**

**React to the
Environment**

Thermostat



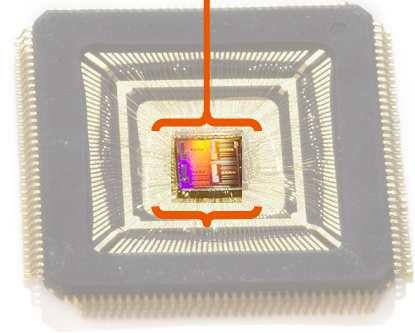
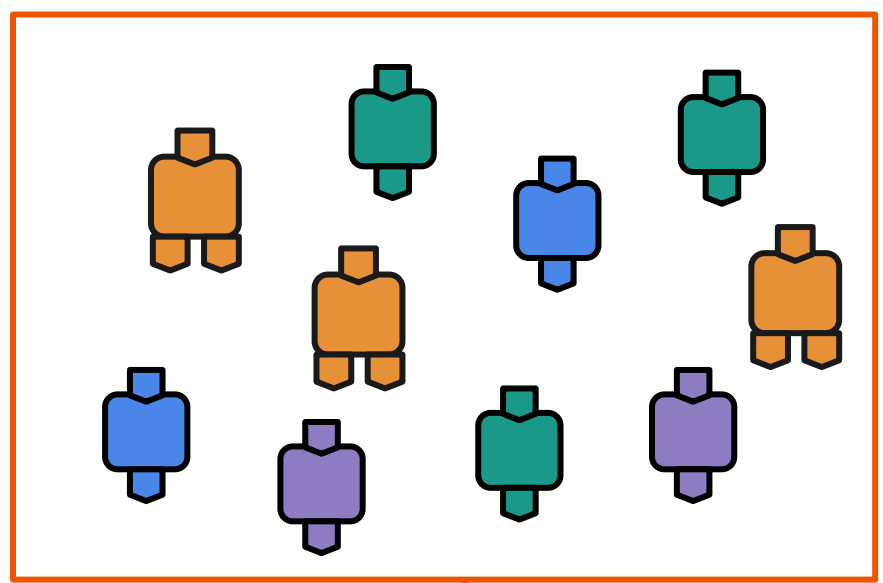
Insulin Pump



**How do these analog devices
work?**

Analog Devices

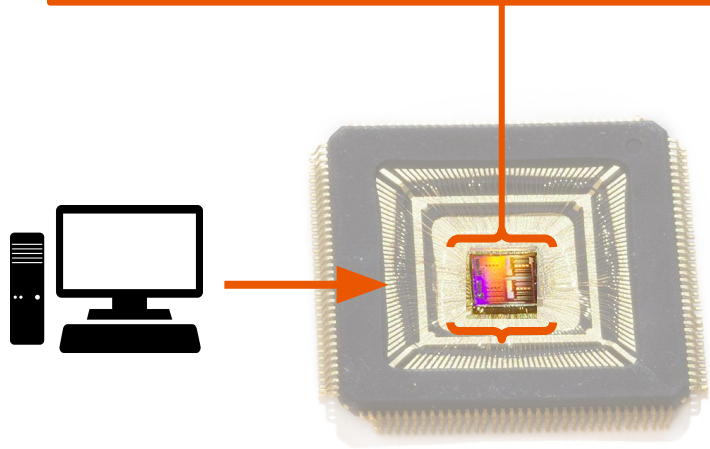
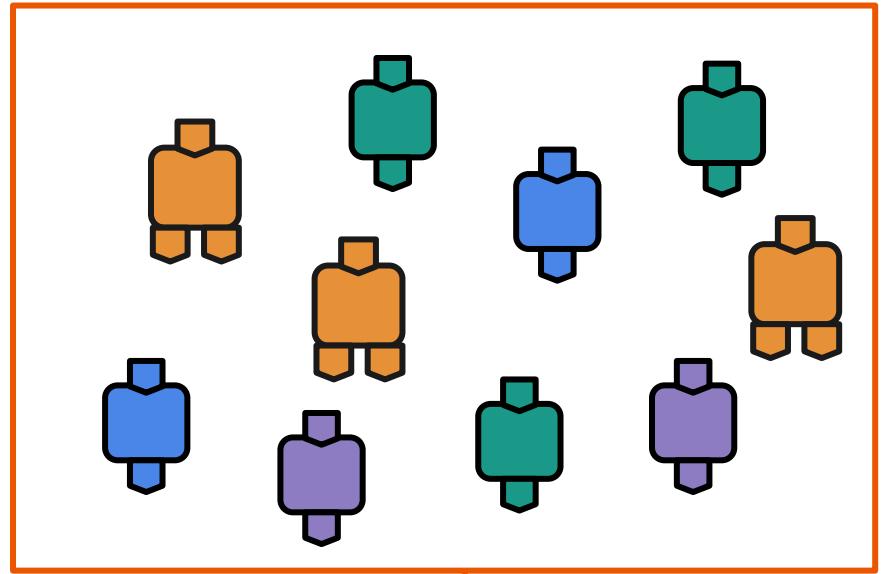
Contain collections
of computational
blocks



Analog Devices



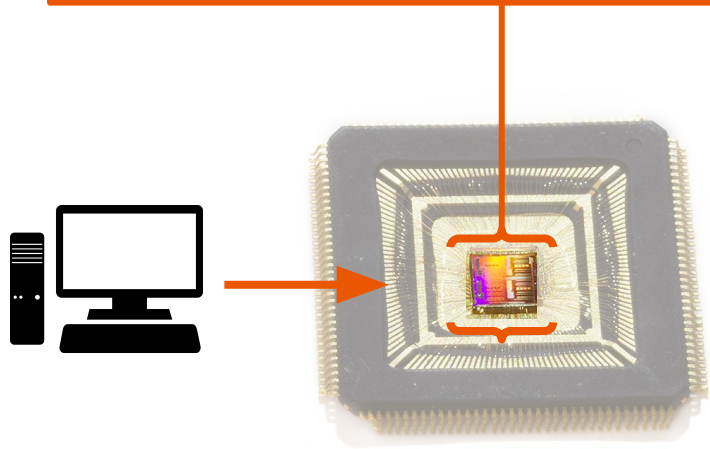
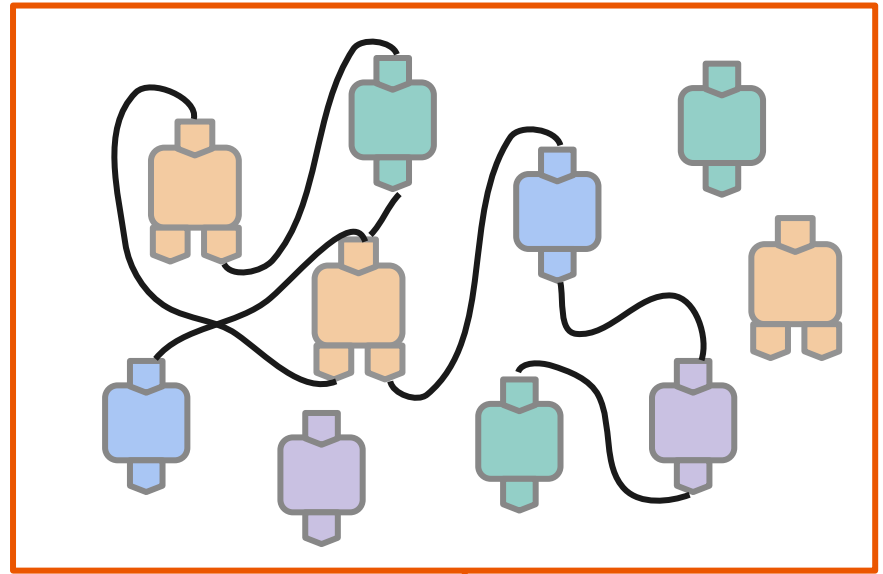
Are programmable



Analog Devices



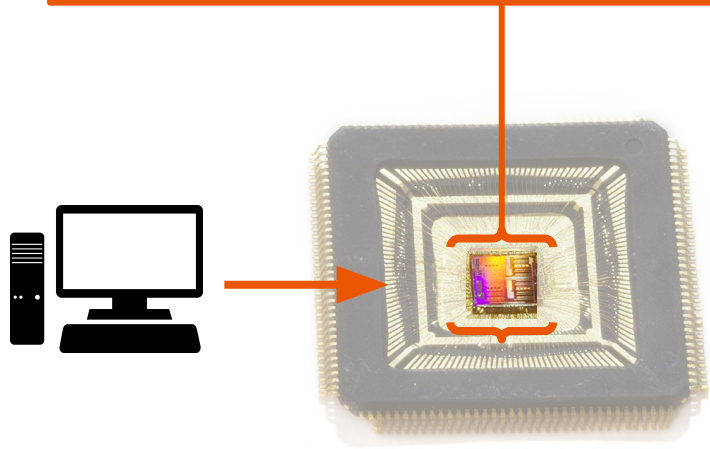
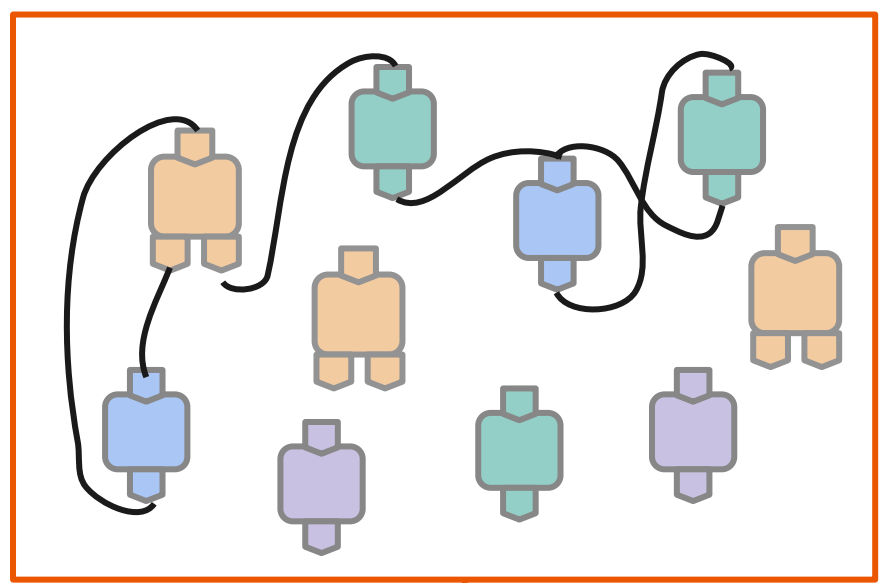
Are programmable



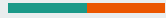
Analog Devices



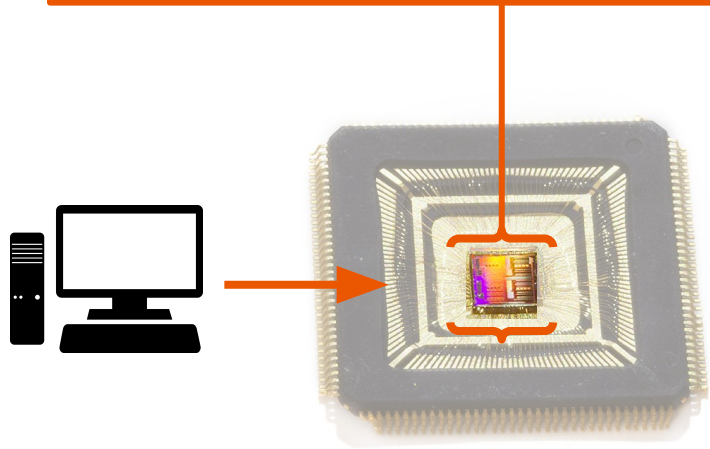
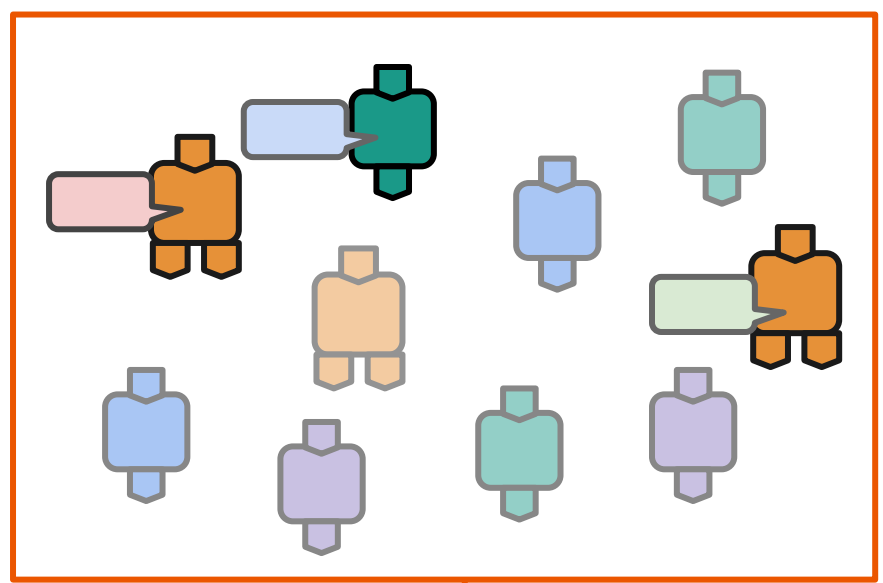
Are programmable



Analog Devices



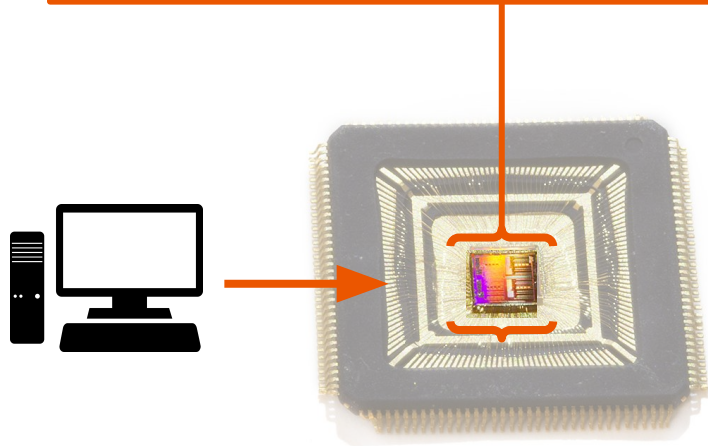
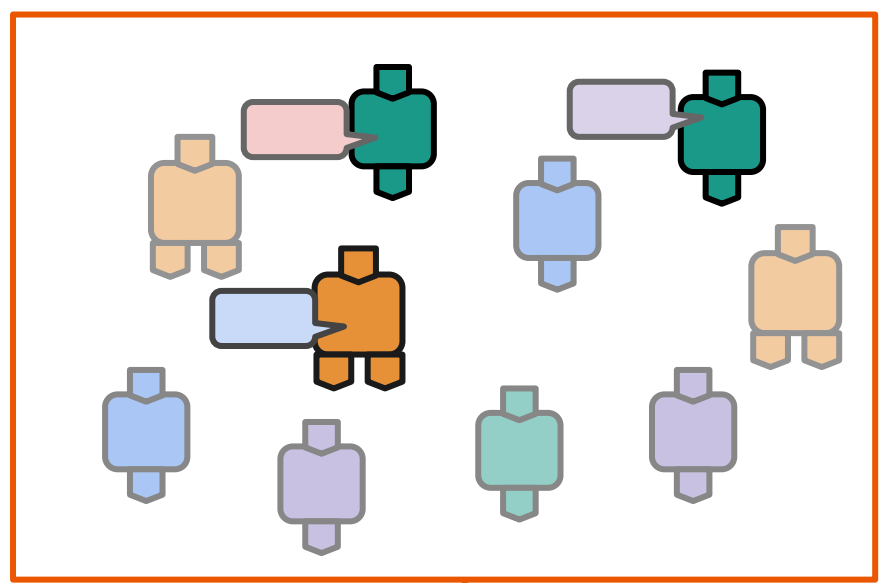
Are programmable



Analog Devices

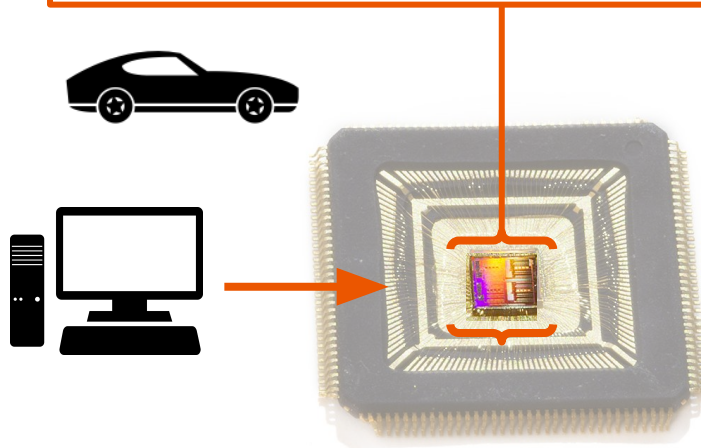
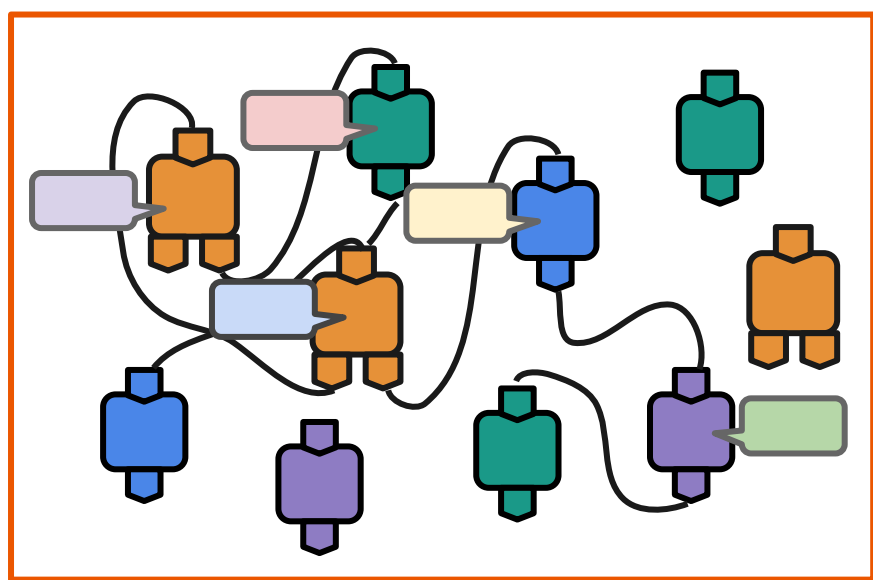


Are programmable



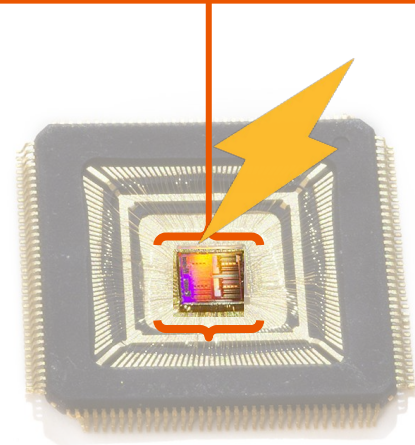
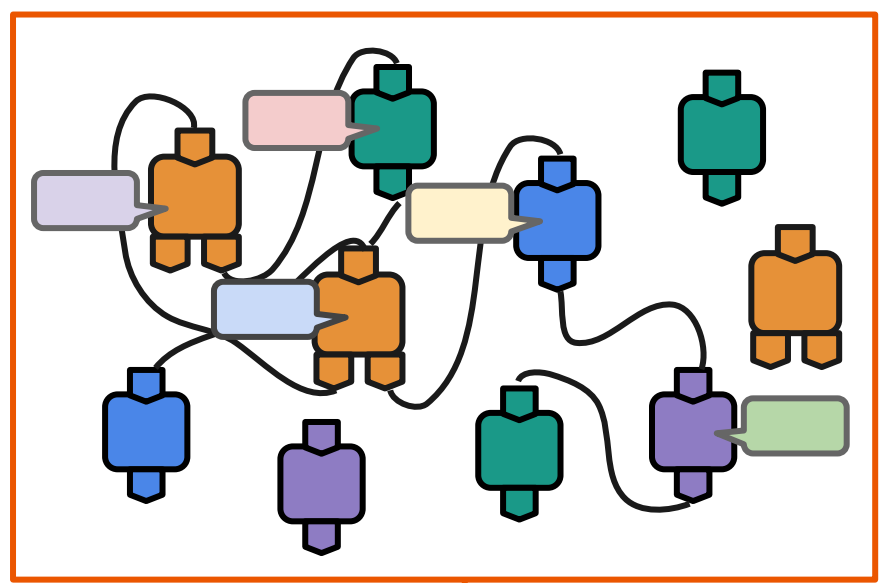
Analog Devices

Perform
computation using
device physics



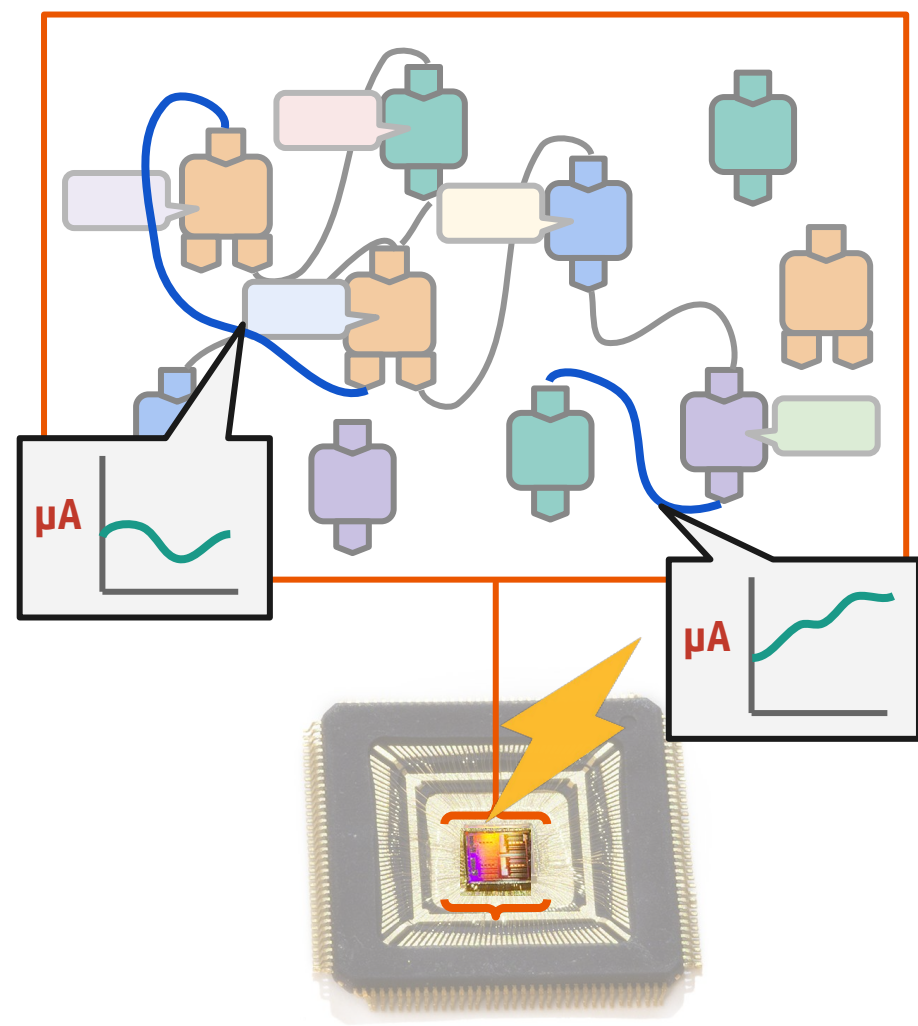
Analog Devices

Perform
computation using
device physics



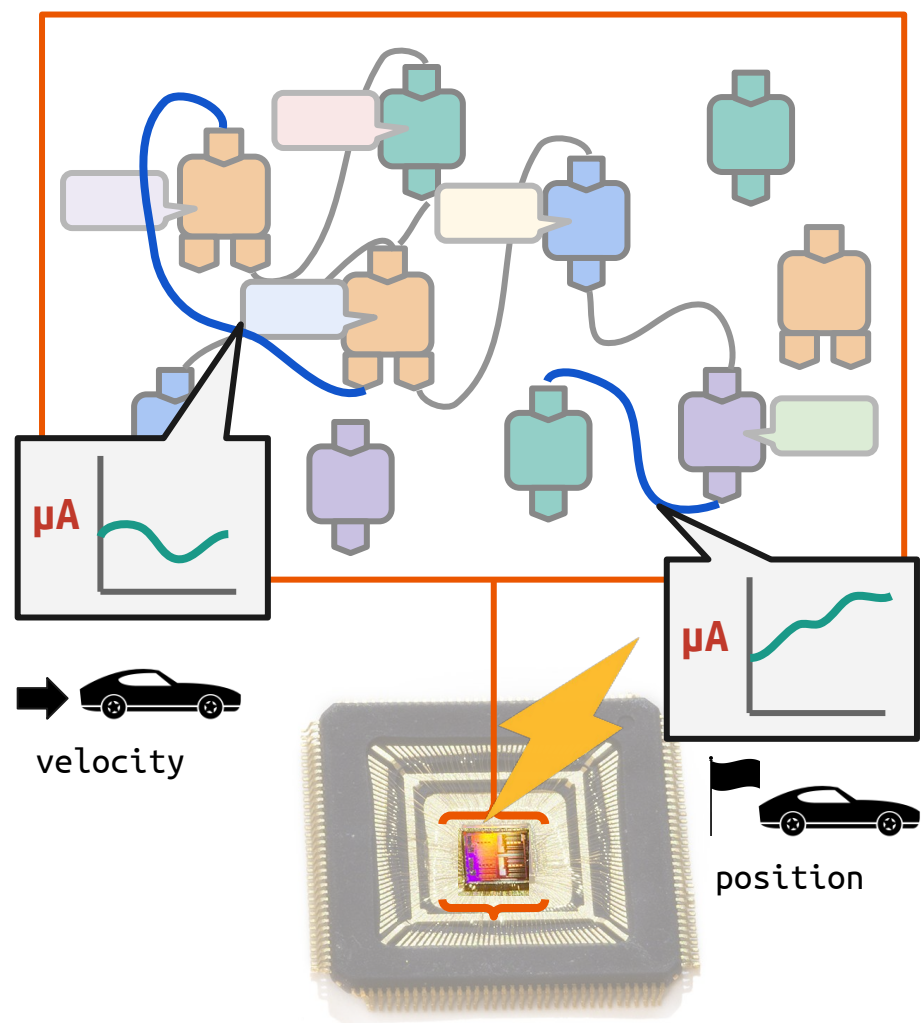
Analog Devices

Perform
computation using
device physics



Analog Devices

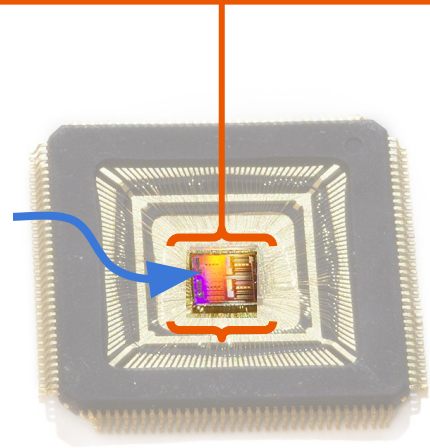
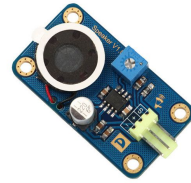
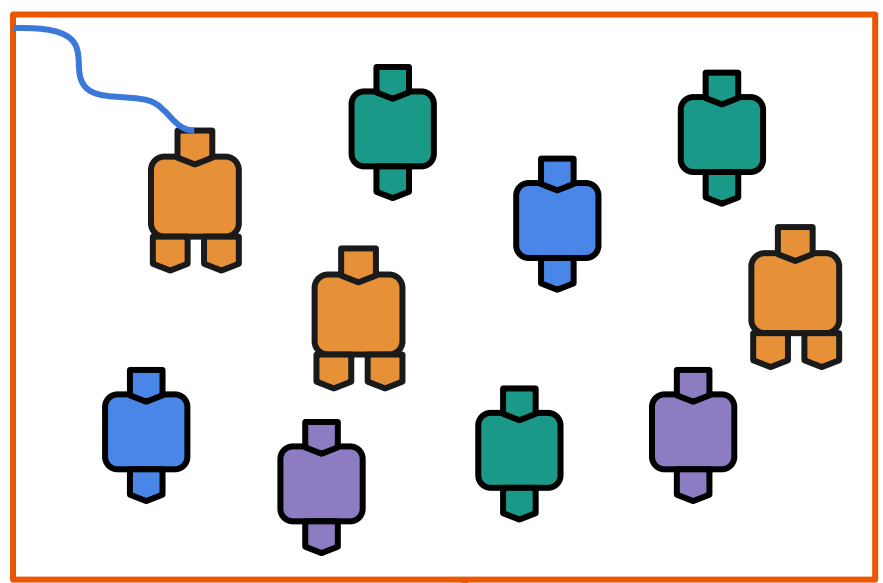
Perform
computation using
device physics



Analog Devices



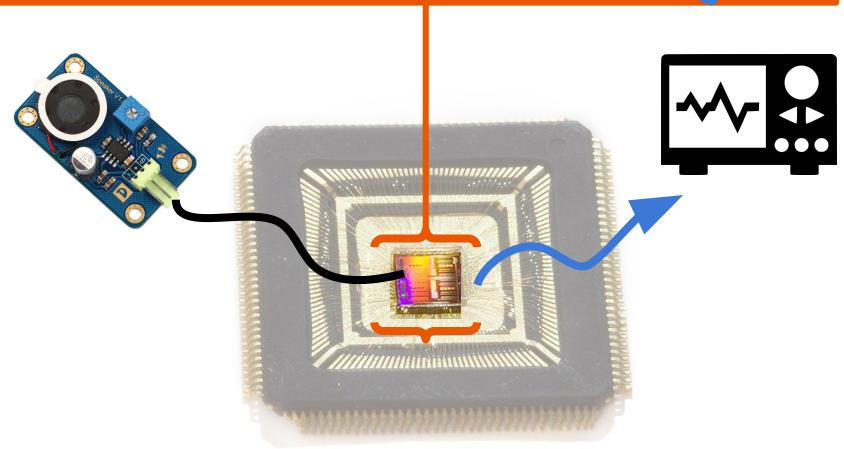
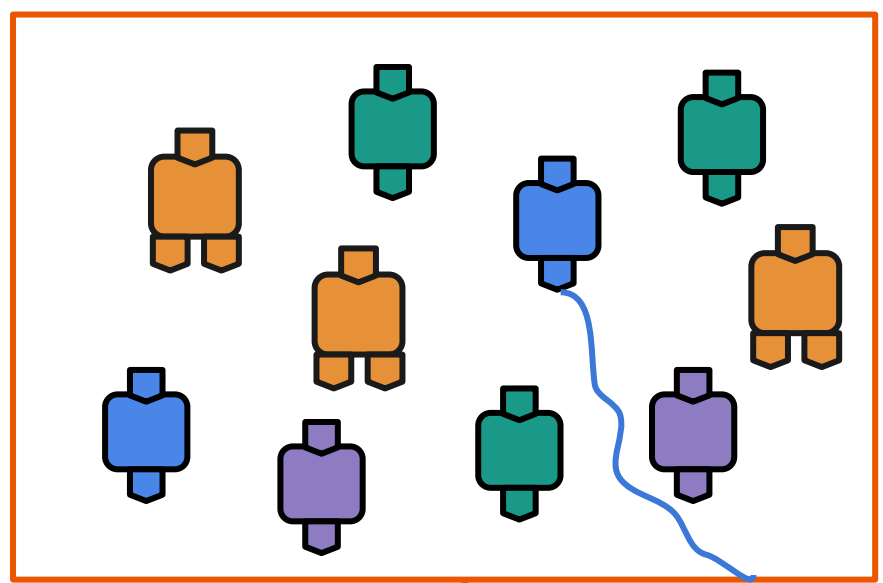
Work with external signals



Analog Devices

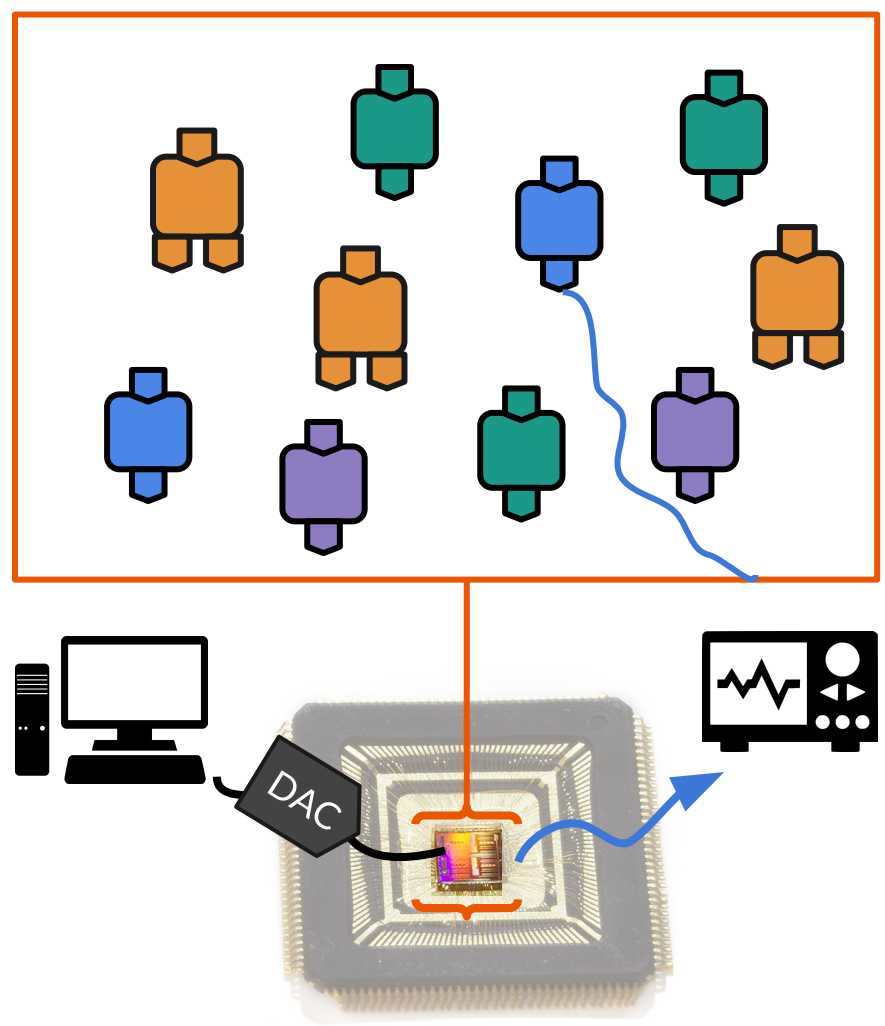


Work with external signals



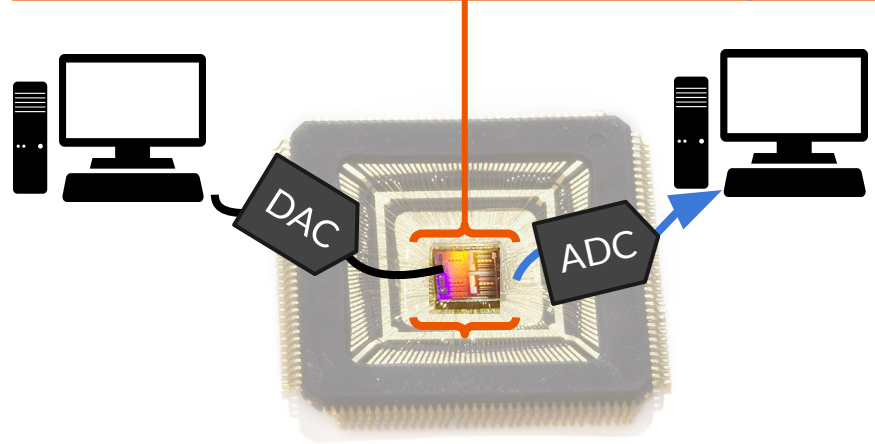
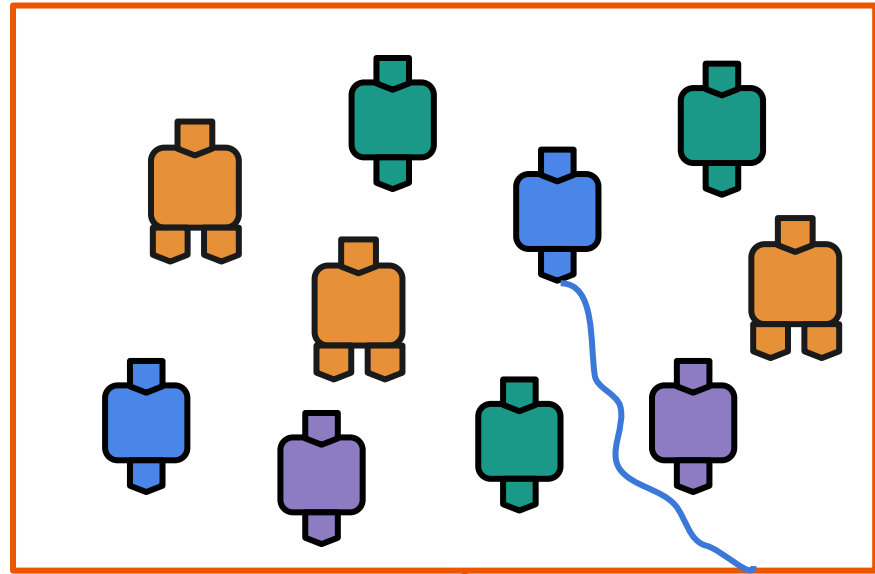
Analog Devices

Work with external signals




Analog Devices

Work with external signals

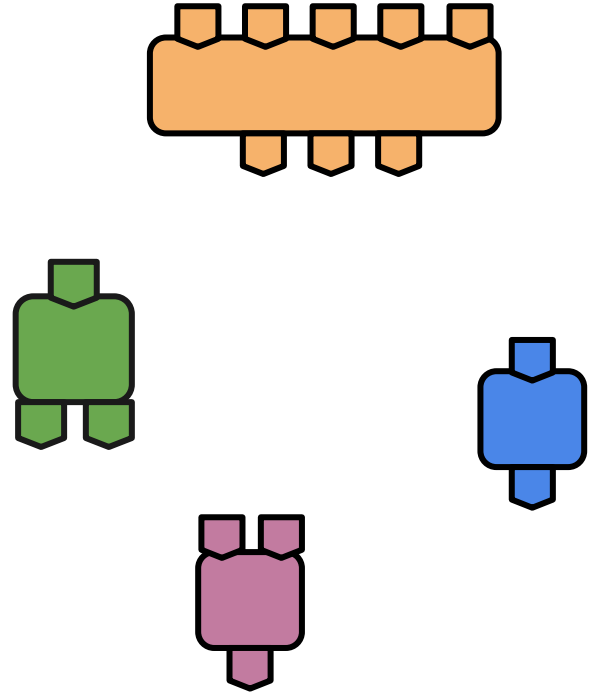


Characteristics of Analog Devices



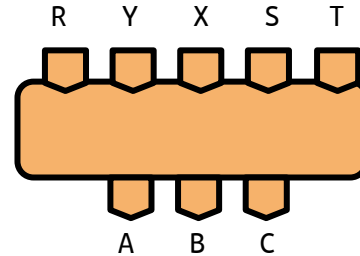


Wide variety of programmable analog blocks

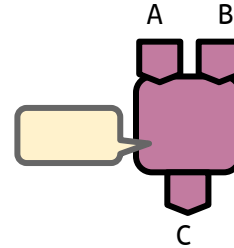


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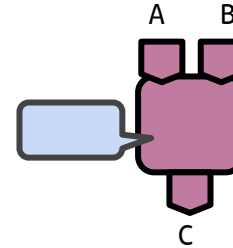
Analog blocks designed for efficiency and for usability



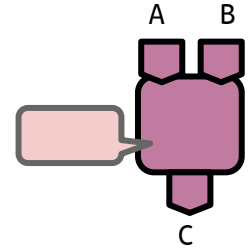
$$A = X - C$$
$$B = Y - C$$
$$dC/dt = \int R * A * B - S * C$$
$$C(0) = T$$



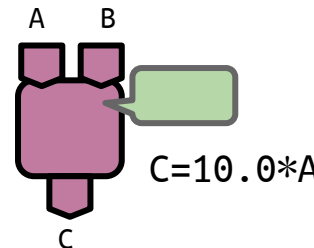
$$C = 0.5 * A * B$$



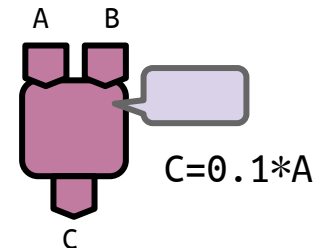
$$C = 0.05 * A * B$$



$$C = 5 * A * B$$



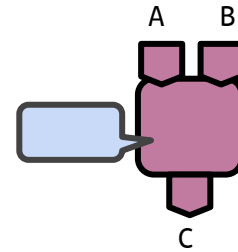
$$C = 10.0 * A$$



$$C = 0.1 * A$$

—

Analog blocks have analog behavior



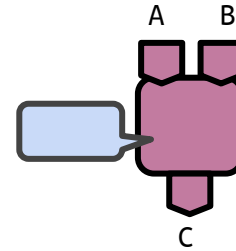
$$C=0.05*A*B$$

—

Analog blocks have analog behavior

Operating Range
Limitations

Frequency
Limitations



Process
Variations

Quantization
Error

$$C=0.05*A*B$$

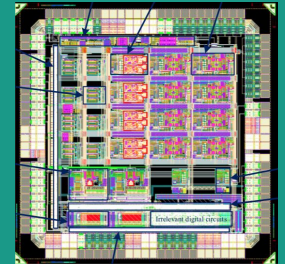
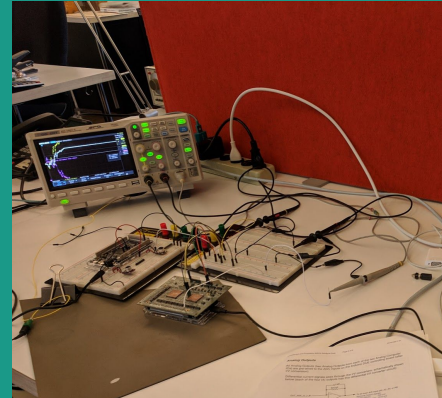
Analog Noise

I developed compilation techniques for this class of devices

1. Time Dilation and Contraction for Programmable Analog Devices with Jaunt. ASPLOS 2018.
2. Configuration Synthesis for Programmable Analog Devices with Arco. PLDI 2016.

I built the first-ever compilation toolchain for a real-world device of this class

Noise-Aware Dynamical System Compilation for Analog Devices with Legno. ASPLOS 2020.



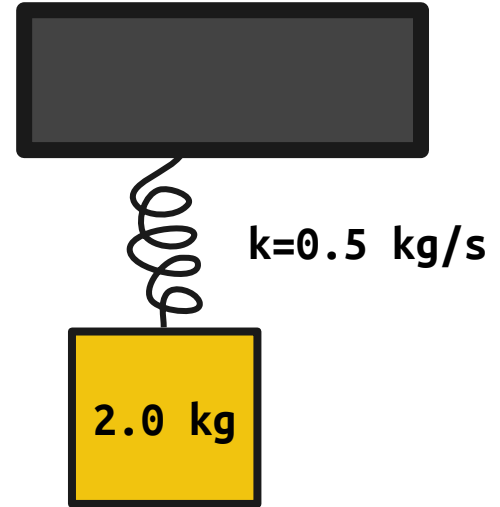
How does the compiler work?

Input



Dynamical System

Harmonic
Oscillator

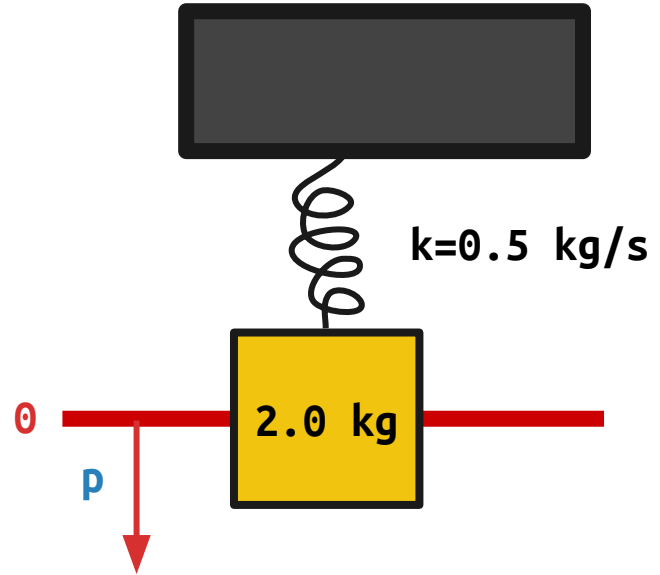


Input



Dynamical System

Harmonic
Oscillator



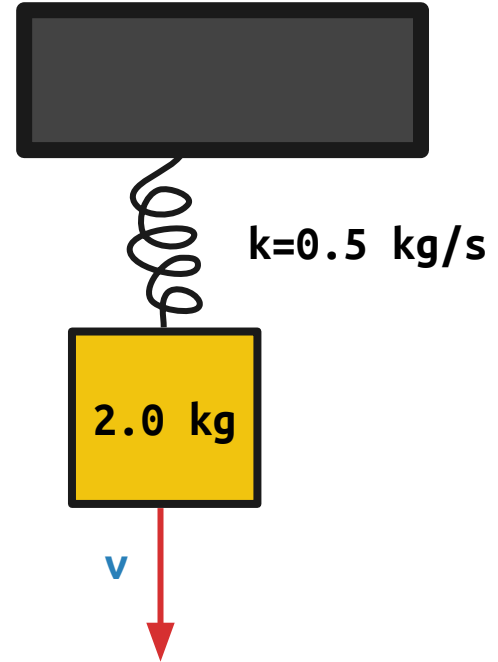
p: position in meters

Input



Dynamical System

Harmonic
Oscillator



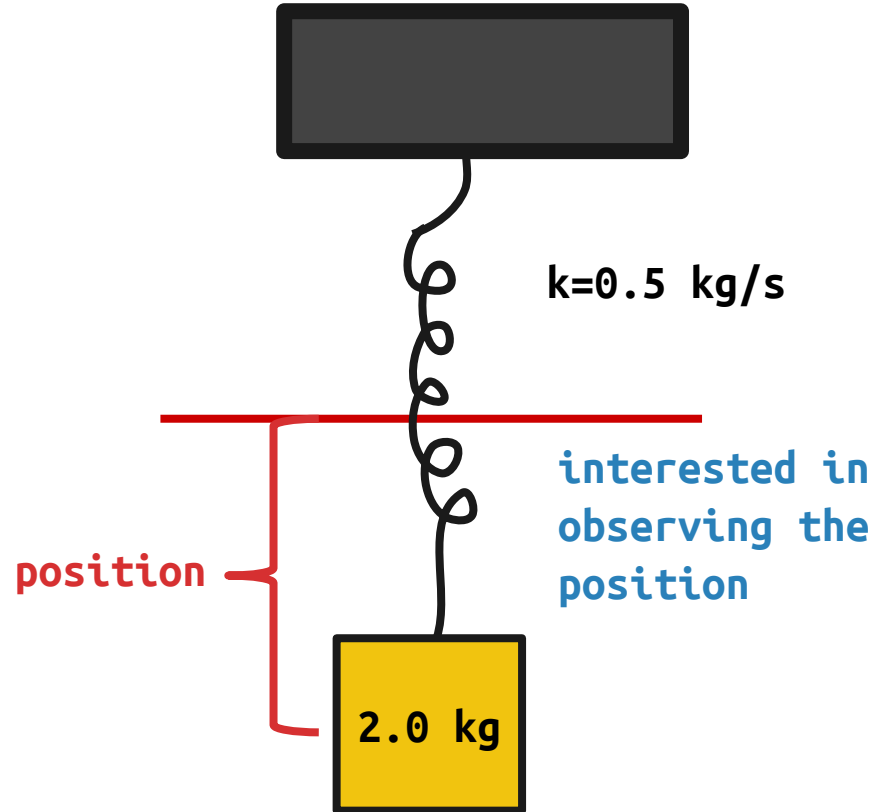
v: velocity in meters/sec

Input



Dynamical System

Harmonic
Oscillator

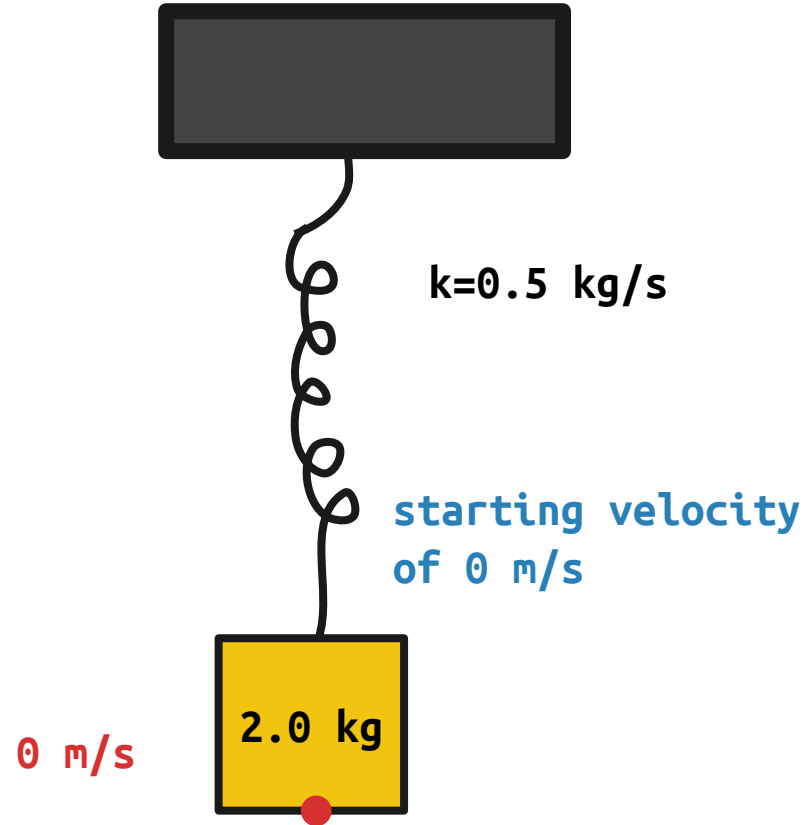


Input



Dynamical System

Harmonic
Oscillator

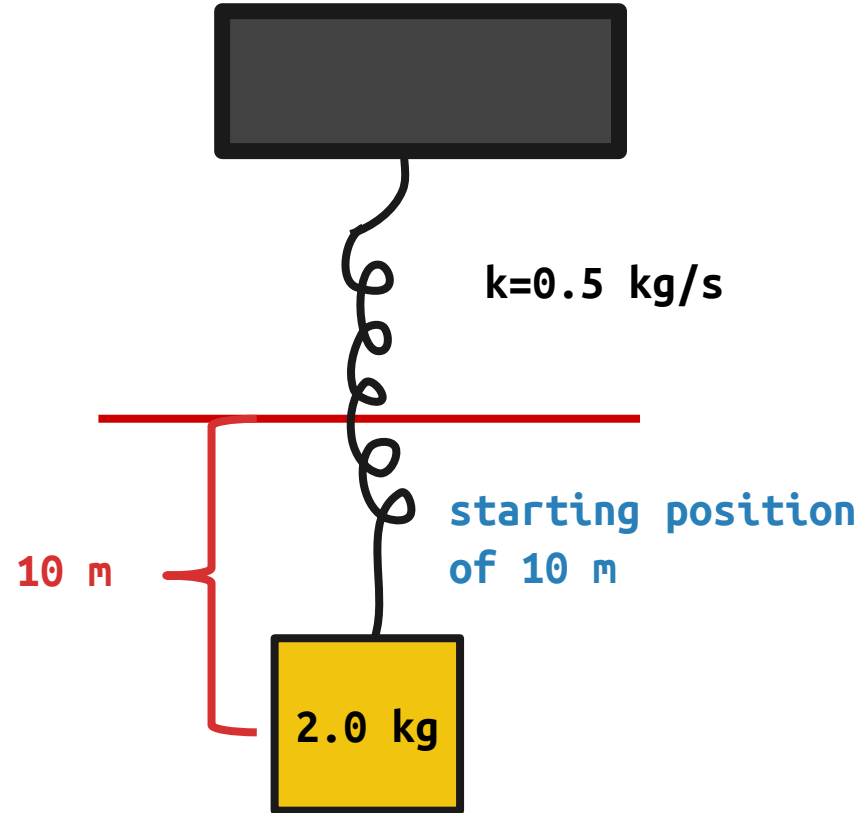


Input



Dynamical System

Harmonic
Oscillator

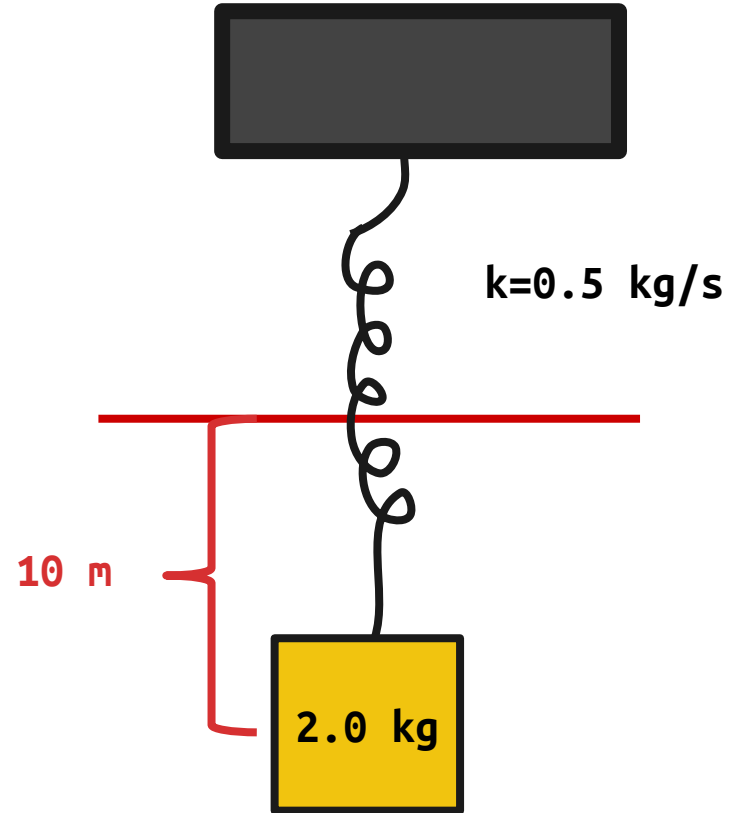
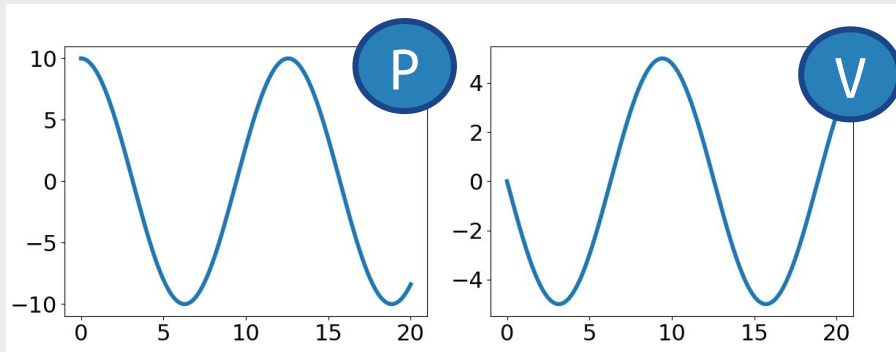


Input



Dynamical System

Harmonic
Oscillator

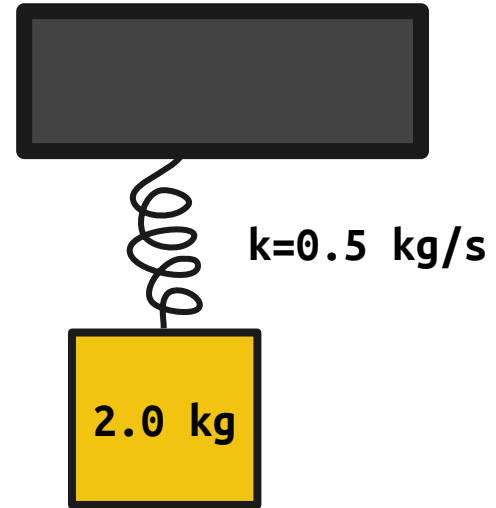


Input



Dynamical System

Harmonic
Oscillator



$$v' = -0.25 * p$$

$$p' = v$$

Input

Dynamical System Program



Oscillator

$$\text{var } v = \int -0.25 * p \, dt_{ds} \quad v(0)=0$$

$$\text{var } p = \int v \, dt_{ds} \quad p(0)=10$$

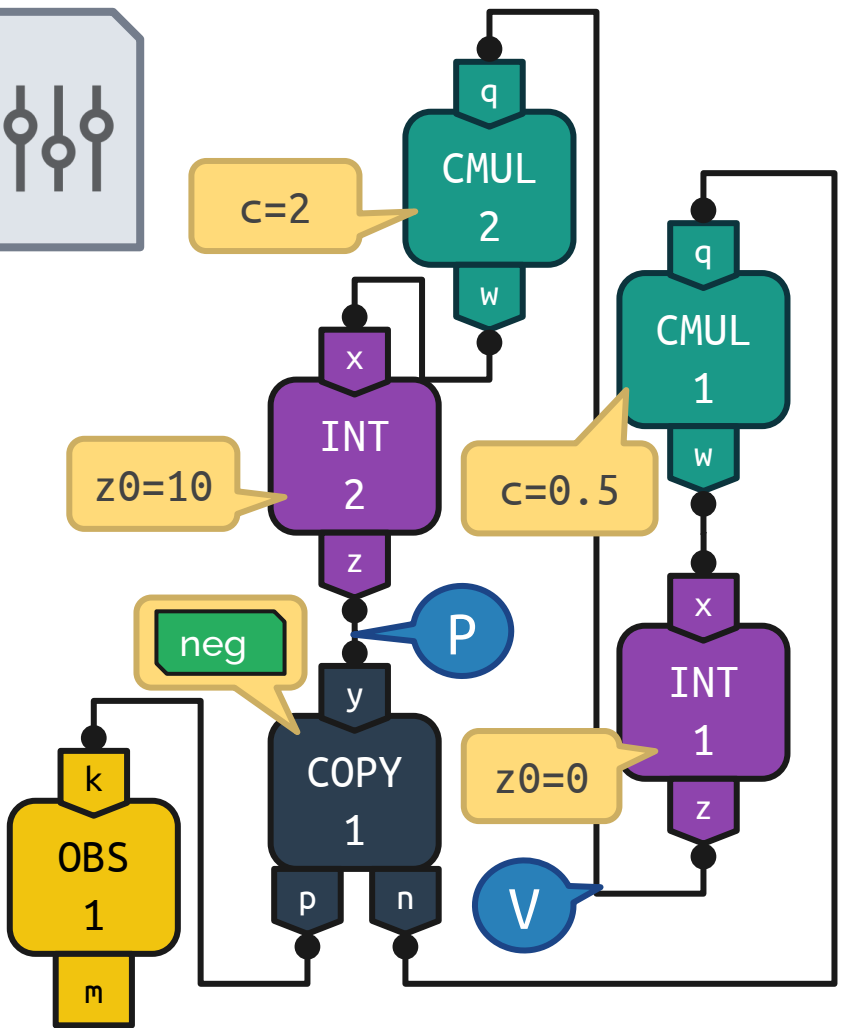
observe p



20 time
units



Oscillator

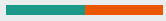


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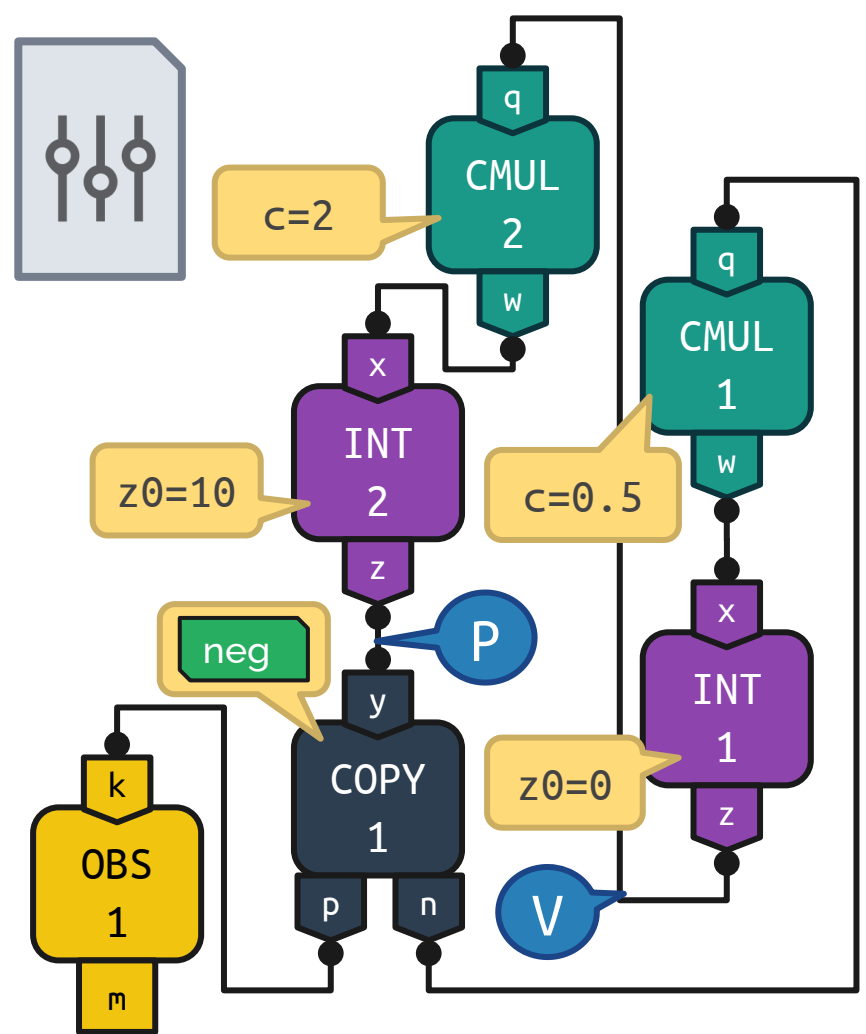
var v = ∫ -0.25*p dt_ds    v(0)=0
var p = ∫ v dt_ds         p(0)=10
observe p

```

Output



Configured Analog Device

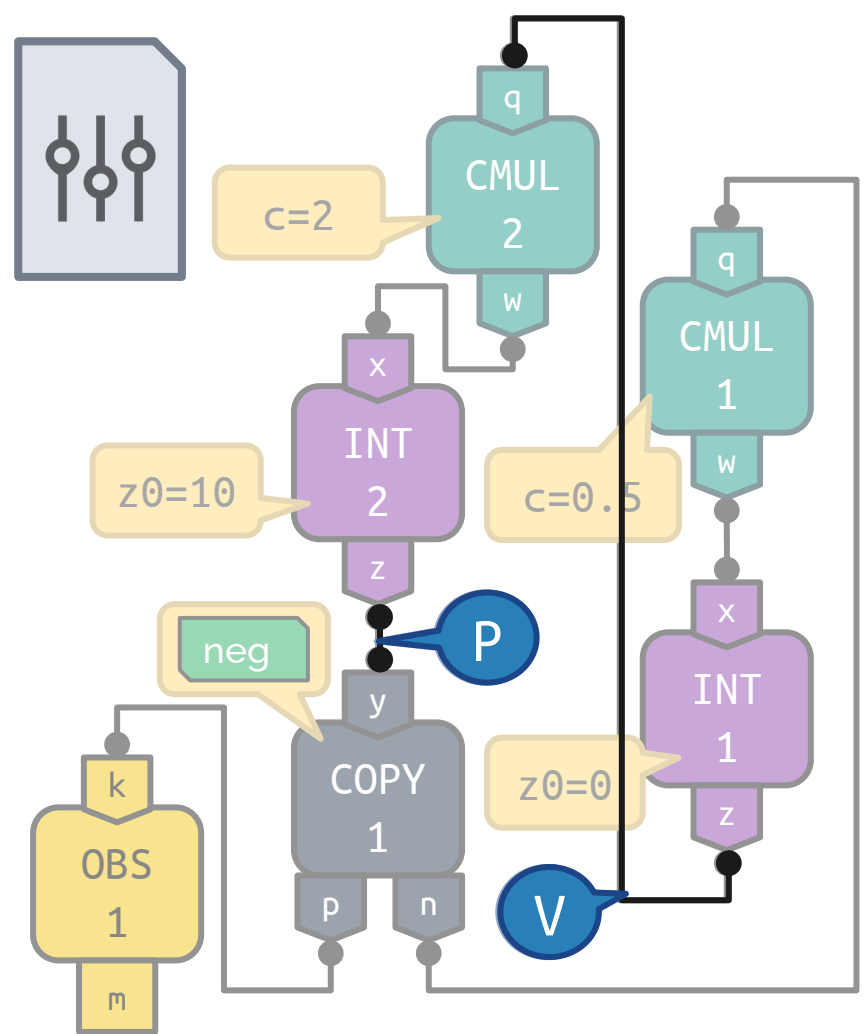


Output



Configured Analog Device

Analog currents map to position and velocity.

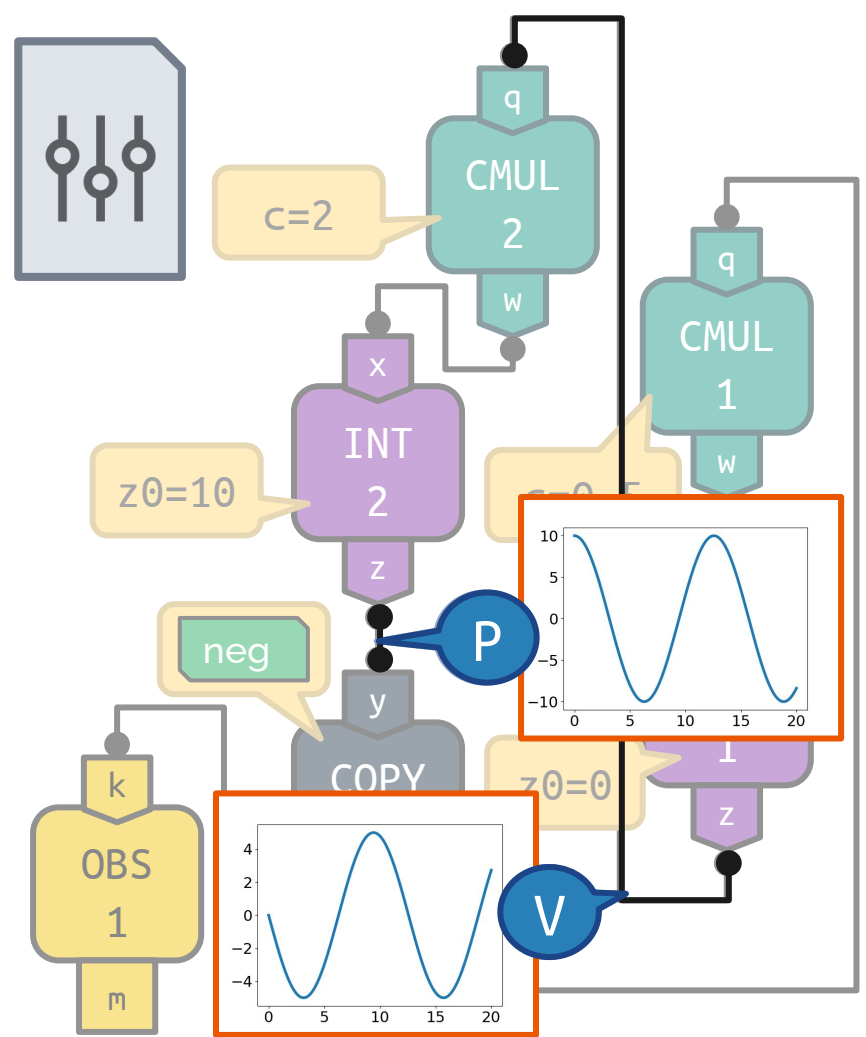


Output



Configured Analog Device

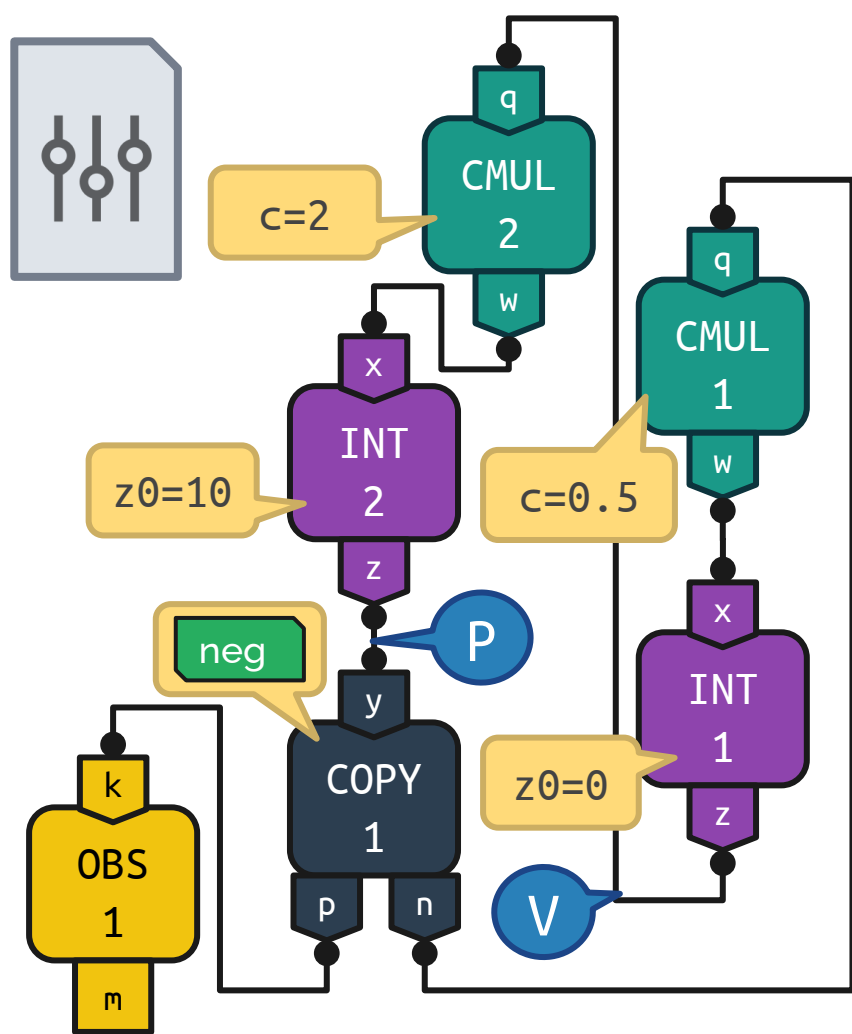
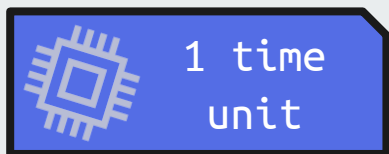
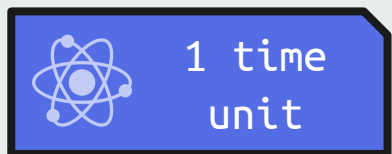
Analog currents map to position and velocity.



Output



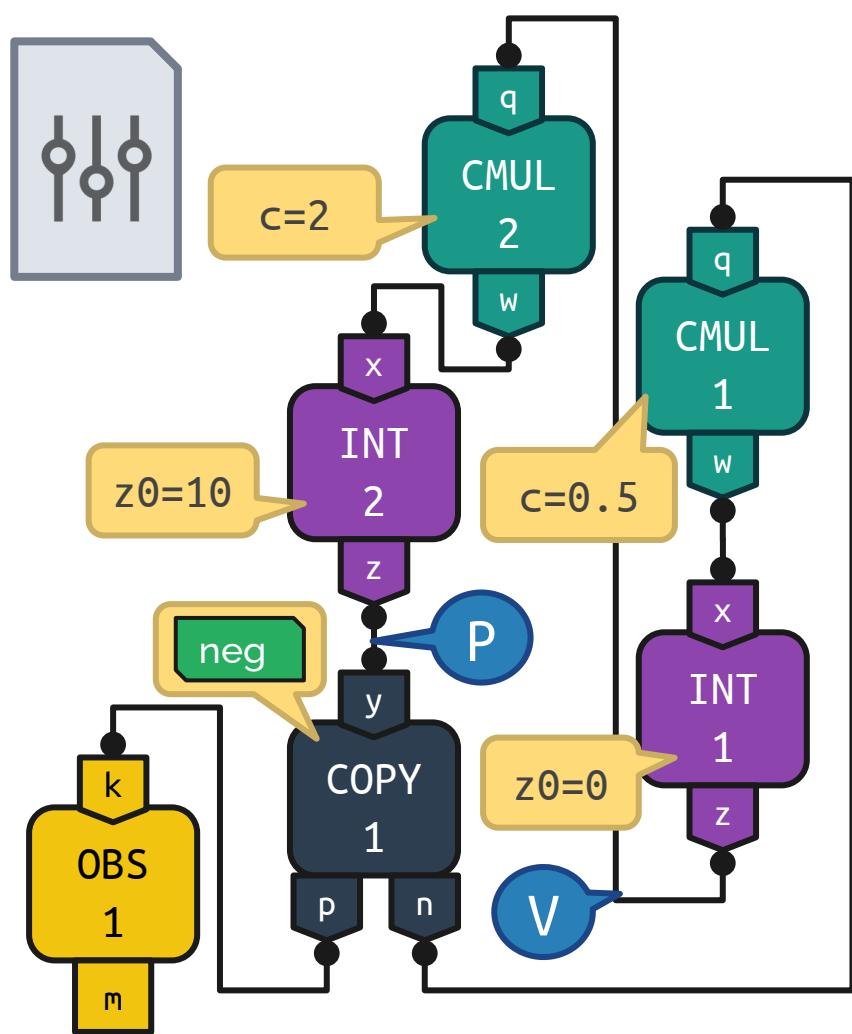
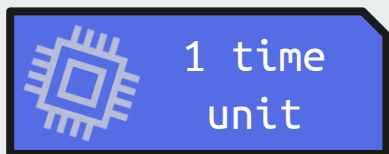
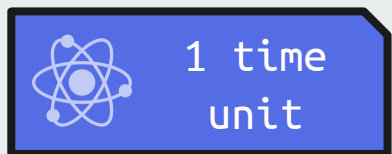
Configured Analog Device



Output




Configured Analog Device




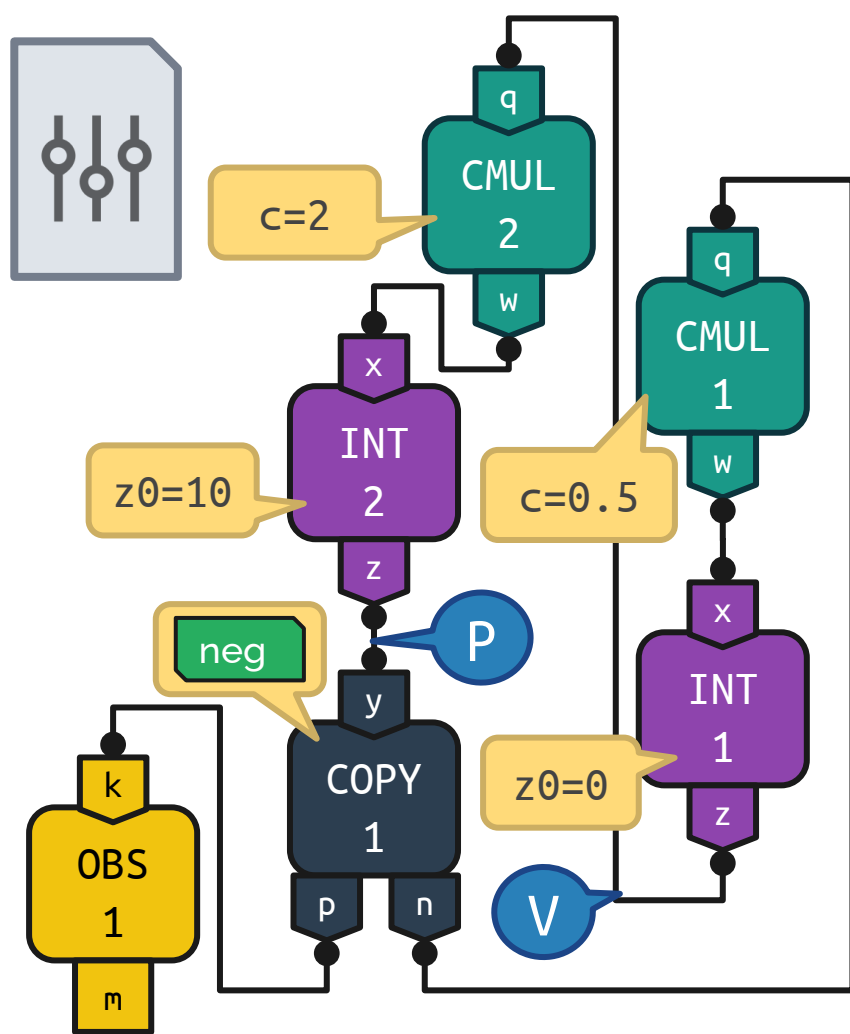
Output



Configured Analog Device

 20 time units

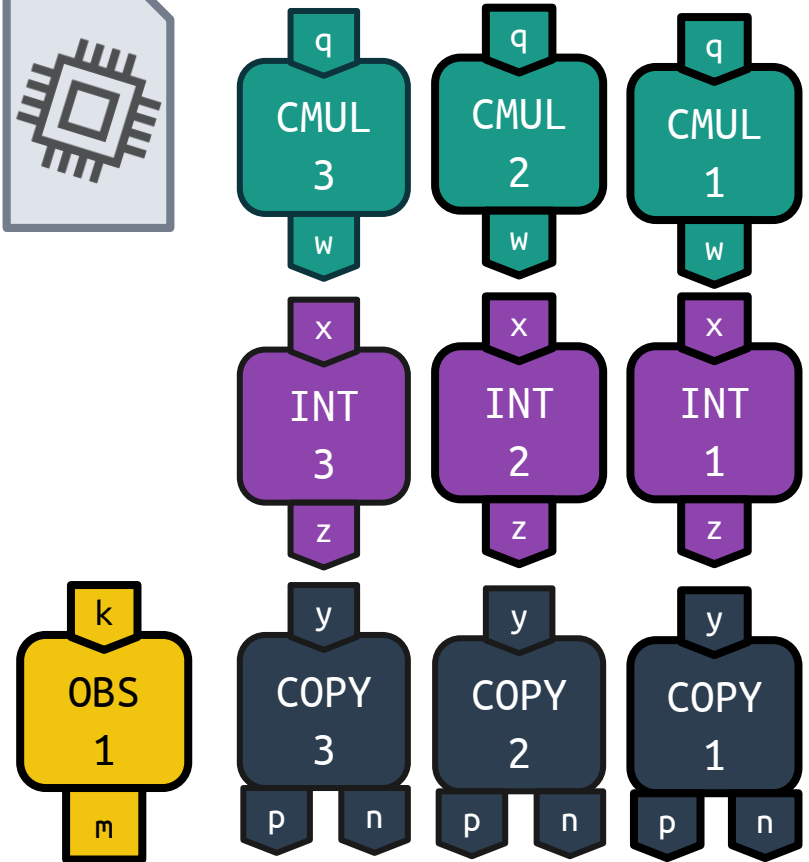
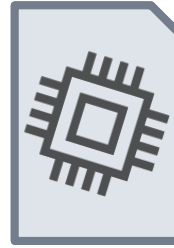
 158.6 μ s



Inputs



Analog Device

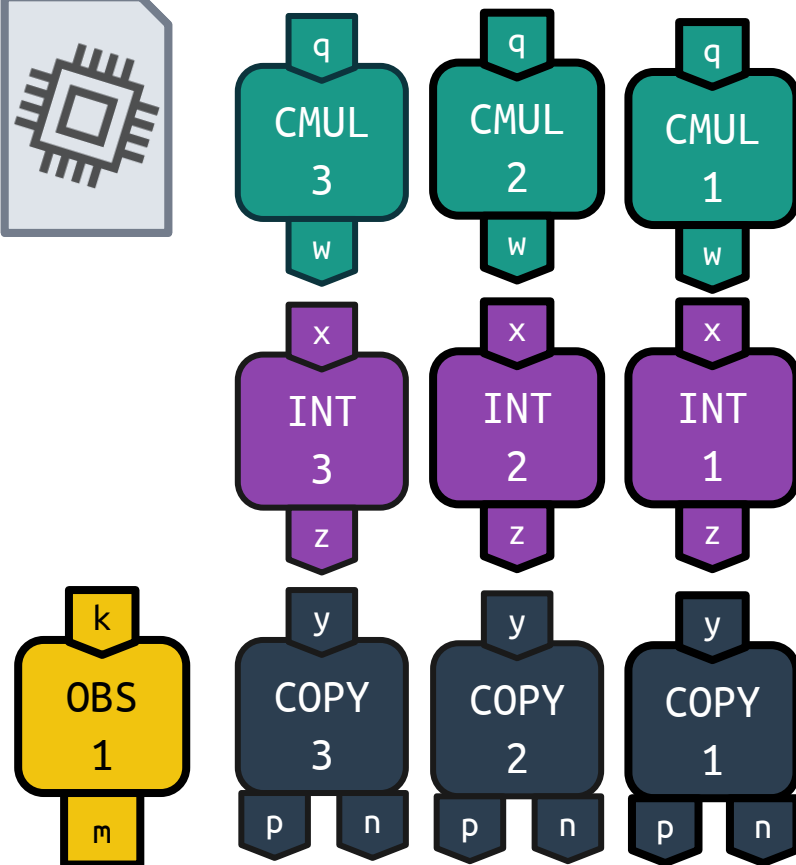
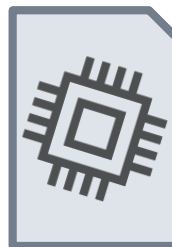


Inputs



Analog Device

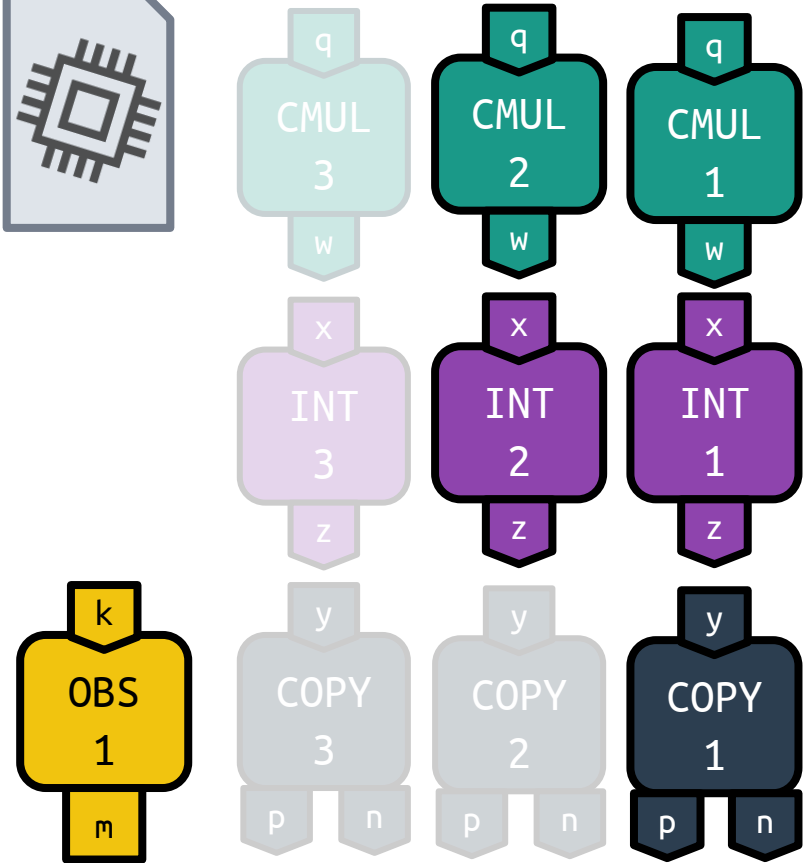
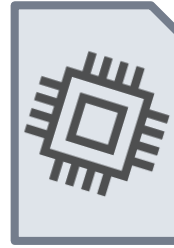
Any block can be connected to any other block.



Inputs



Analog Device

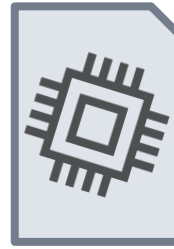


Inputs

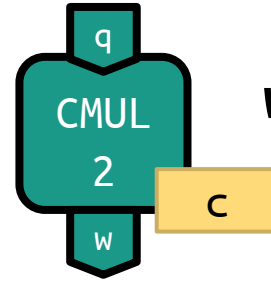


Programming Interface

Digitally settable constants

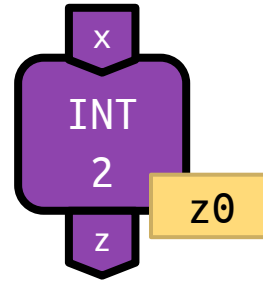


Multiplier



$$w = c * q$$

Integrator



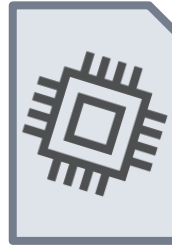
$$z = \int 0.5 * x \, dt_{hw}$$
$$z(0) = z_0$$

Inputs



Programming Interface

Digitally settable block modes



Digitally Settable
Block Modes



$$p = y$$

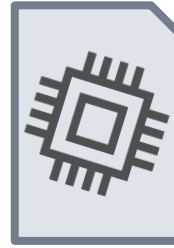
$$n = y$$

Inputs



Programming Interface

Digitally settable block modes



Digitally Settable
Block Modes

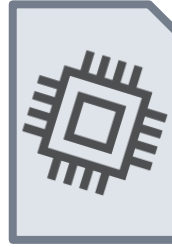


$$p = x$$
$$n = -x$$

Inputs



Observing Signals

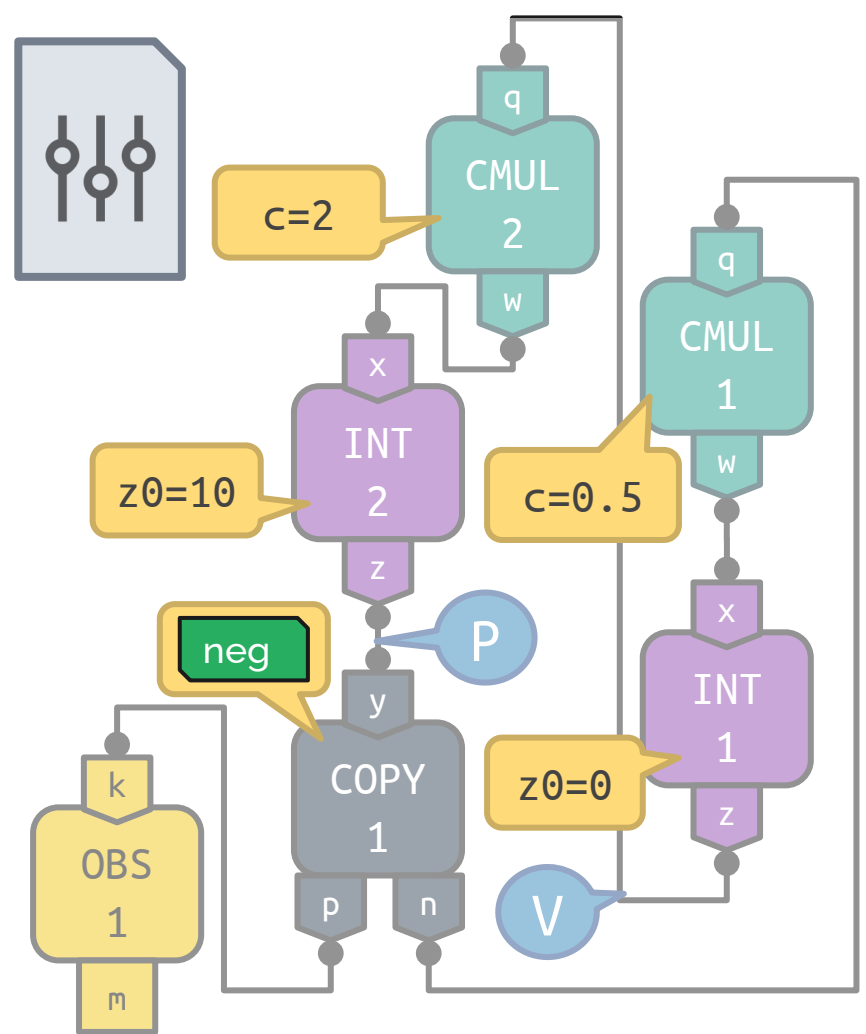


Output



Analog Device Configuration

Programs each block in configuration

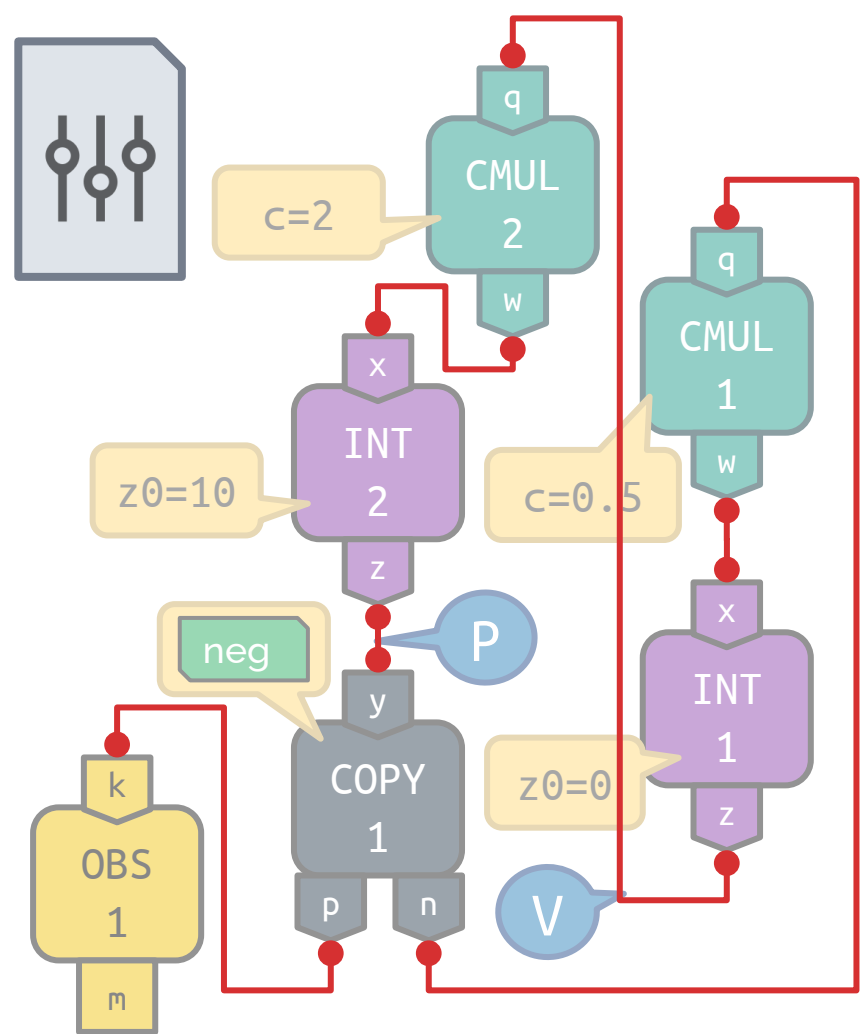


Output

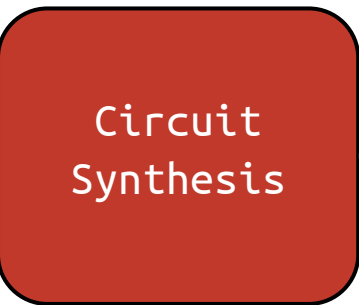
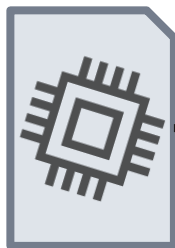


Configured Analog Device

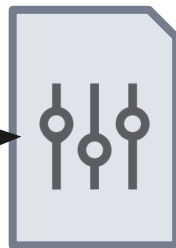
Selects connections to enable on device.



Dynamical System
Specification

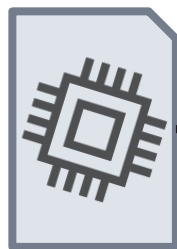
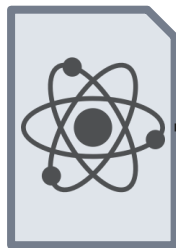


Analog Device
Configuration

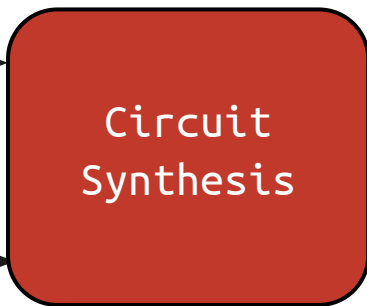


Analog Device
Specification

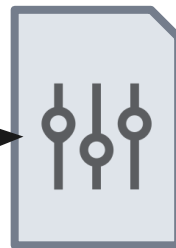
Dynamical System
Specification



Analog Device
Specification



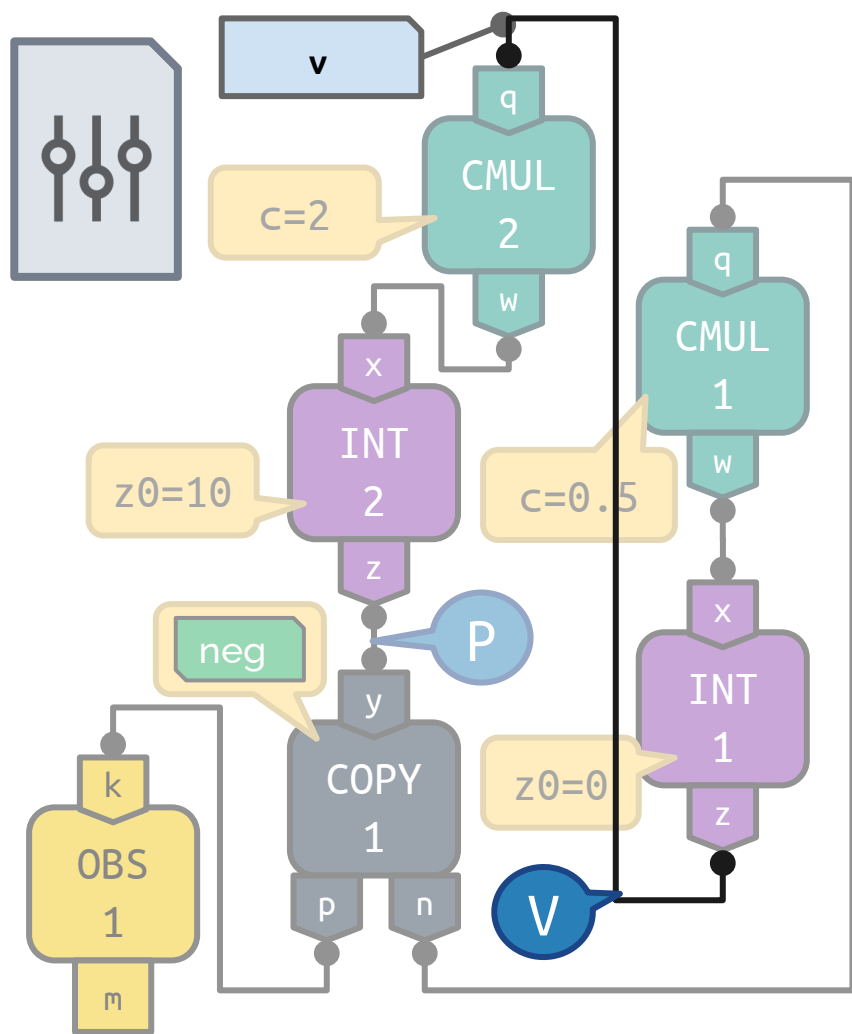
Analog Device
Configuration



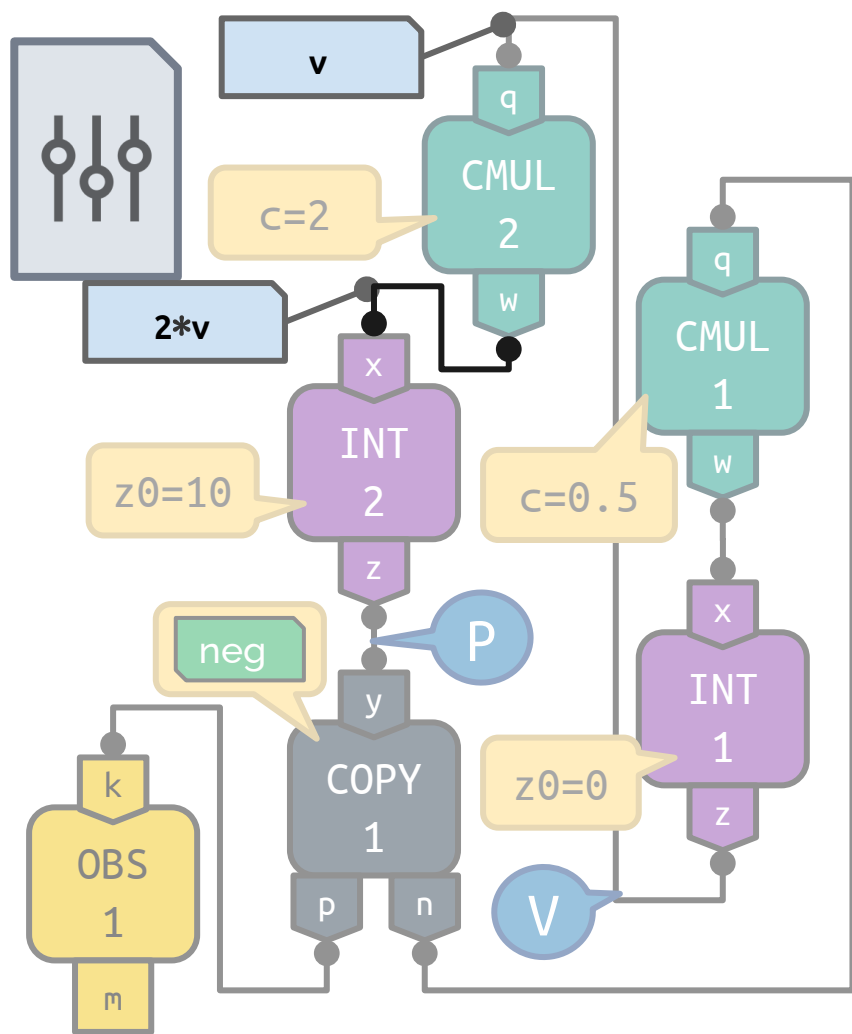
Dynamics that
match



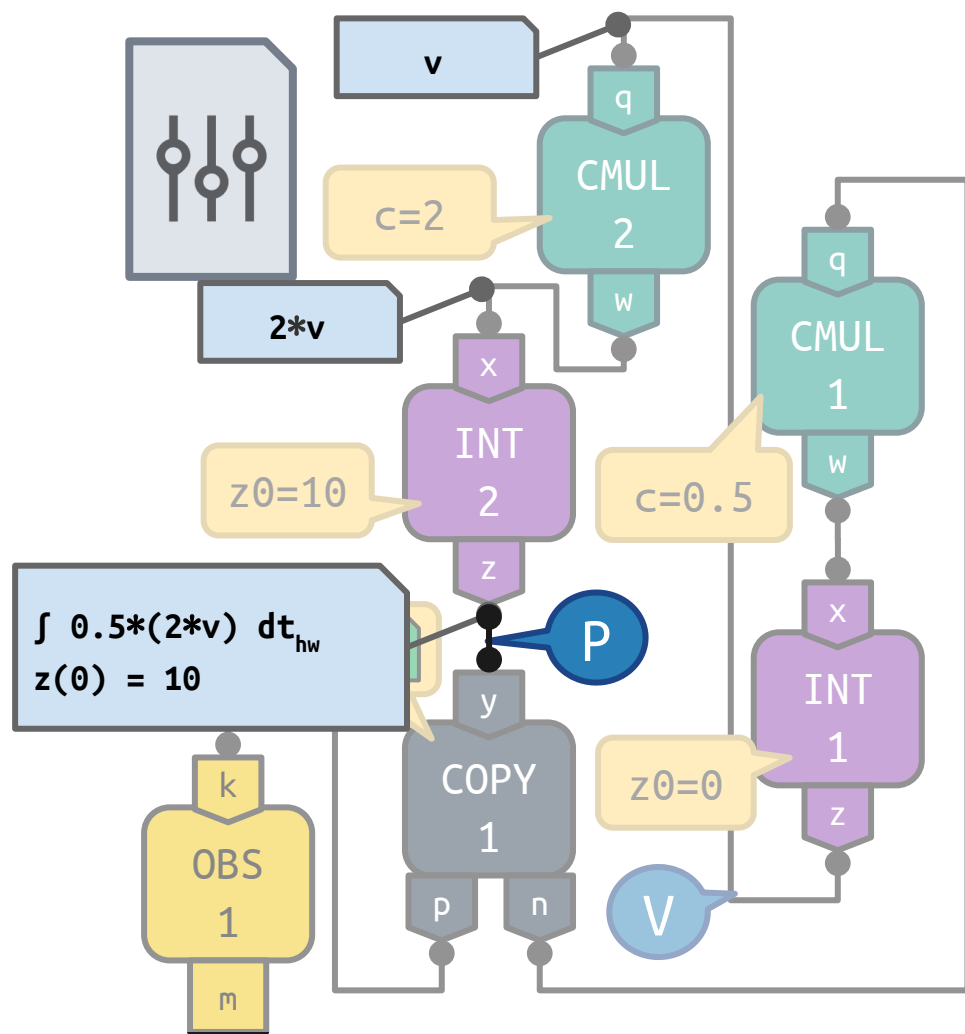
Matching Dynamics



Matching Dynamics



Matching Dynamics

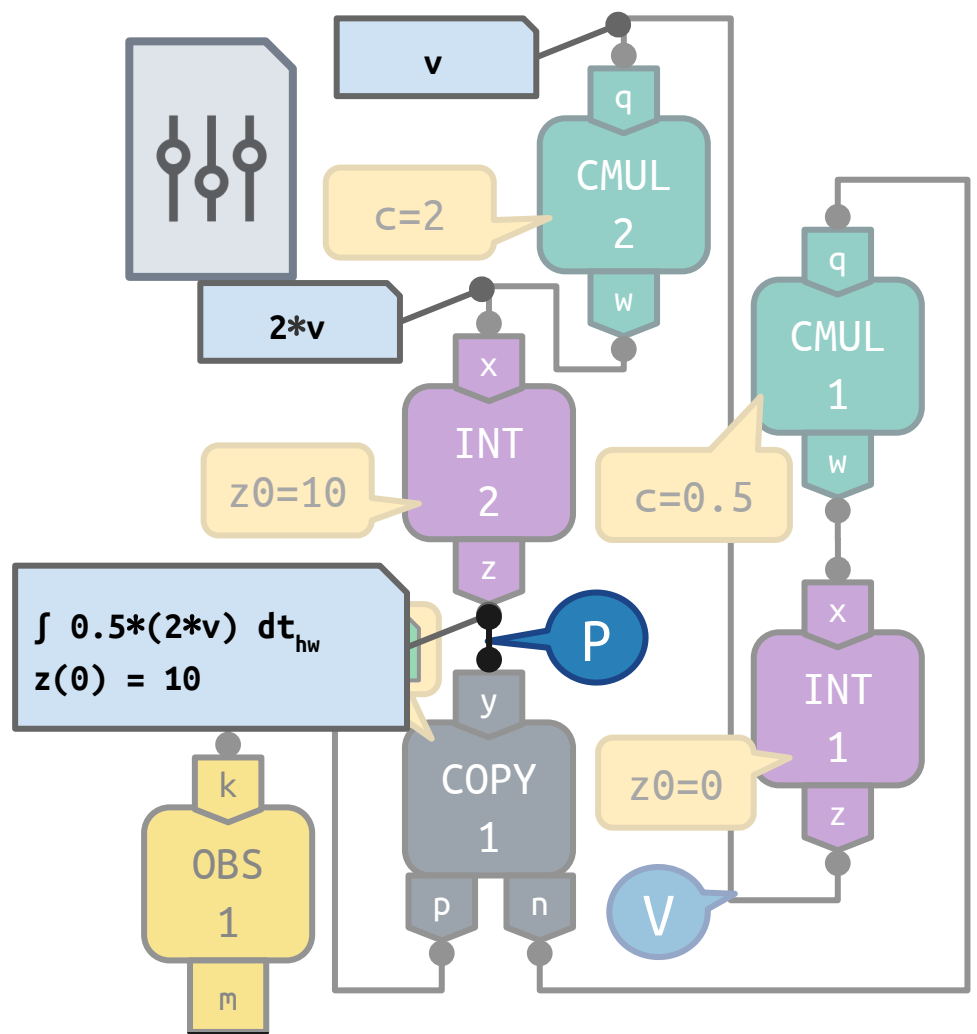


Matching Dynamics

P

$$p = \int 0.5 * (2 * v) dt_{hw}$$

$$p(0) = 10$$



Matching Dynamics

P

$$p = \int 0.5 * (2 * v) dt_{hw}$$

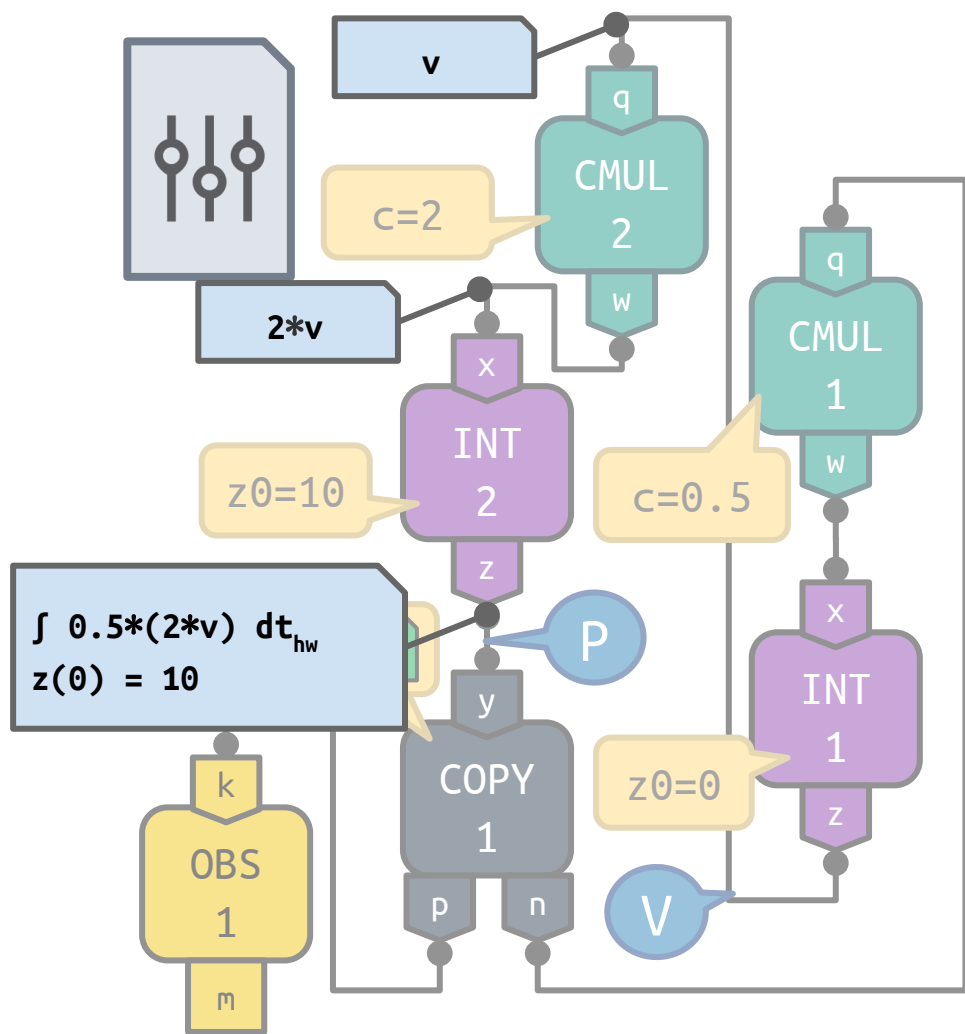
$$p(0) = 10$$

==



$$\text{var } p = \int v dt_{ds}$$

$$p(0) = 10$$



Matching Dynamics

P $p = \int 0.5*(2*v) dt_{hw}$
 $p(0)=10$

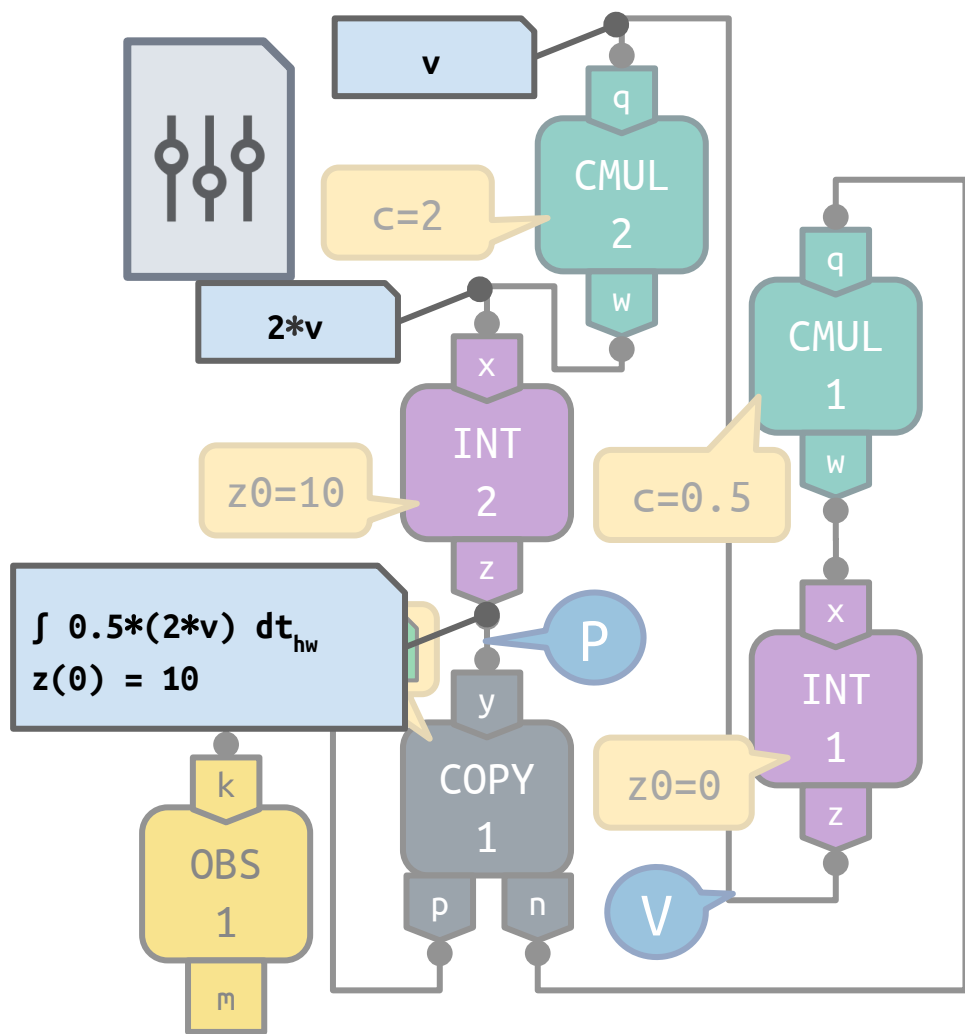


var $p = \int v dt_{ds}$
 $p(0)=10$

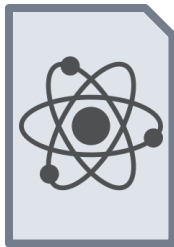
V $v = \int 0.5*(0.5*(-p)) dt_{hw}$
 $v(0)=0$



var $v = \int -0.25*p dt_{ds}$
 $v(0)=0$



Circuit Synthesis



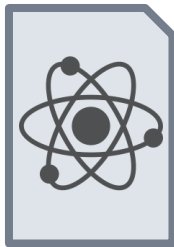
```
var v = ∫ -0.25*p dtds    v(0)=0  
var p = ∫ v dtds        p(0)=10  
observe p
```

Circuit Synthesis

Subcircuit
Synthesis

Assembly

Place
and Route



var v = $\int -0.25 * p \, dt_{ds}$ v(0)=0

observe p

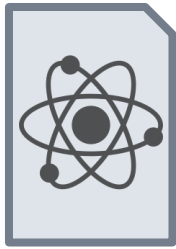
var p = $\int v \, dt_{ds}$ p(0)=10

Circuit Synthesis

Subcircuit
Synthesis

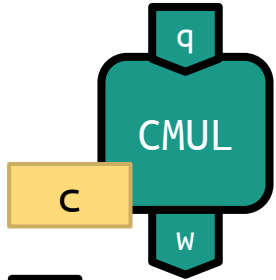
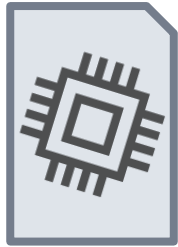
Assembly

Place
and Route

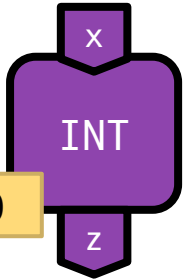


var v = $\int -0.25 * p \, dt_{ds}$ v(0)=0

observe p



w = c * q



z = $\int 0.5 * x \, dt_{hw}$
z(0) = z0

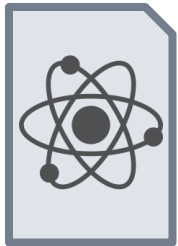
var p = $\int v \, dt_{ds}$ p(0)=10

Circuit Synthesis

Subcircuit
Synthesis

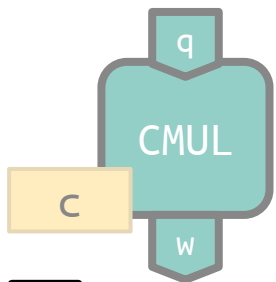
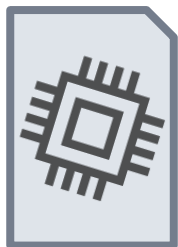
Assembly

Place
and Route

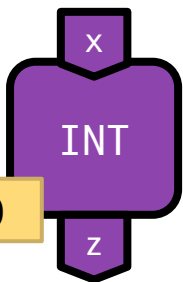


var v = $\int -0.25 * p \ dt_{ds}$ v(0)=0

observe p

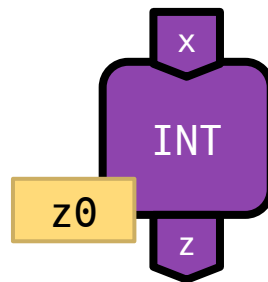


$w = c * q$



$z = \int 0.5 * x \ dt_{hw}$
z(0)=z0

Block Selection



$z = \int 0.5 * x \ dt_{hw}$
z(0)=z0

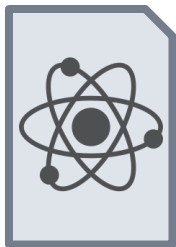
var p = $\int v \ dt_{ds}$ p(0)=10

Circuit Synthesis

Subcircuit
Synthesis

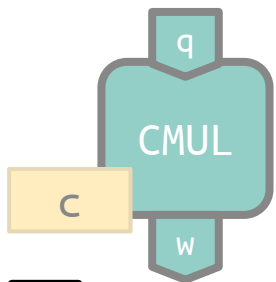
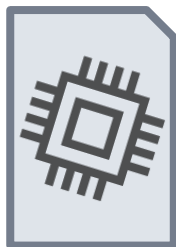
Assembly

Place
and Route

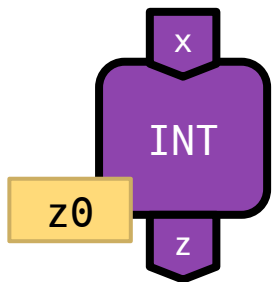


var v = $\int -0.25 * p \ dt_{ds}$ v(0)=0

observe p

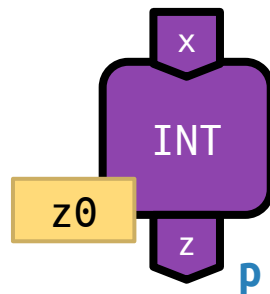


w = c*q



$z = \int 0.5 * x \ dt_{hw}$
z(0)=z0

Unification



$z = \int 0.5 * x \ dt_{hw}$
z(0)=z0

var p = $\int v \ dt_{ds}$ p(0)=10

Goal

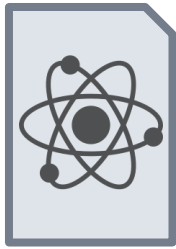
z = p

Circuit Synthesis

Subcircuit
Synthesis

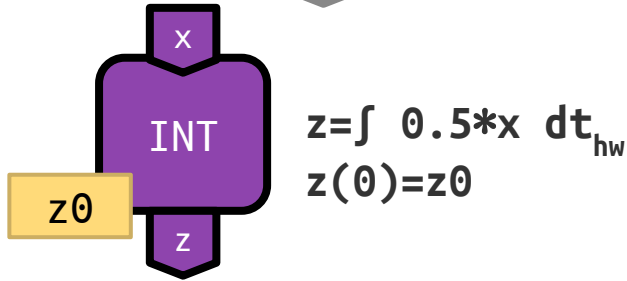
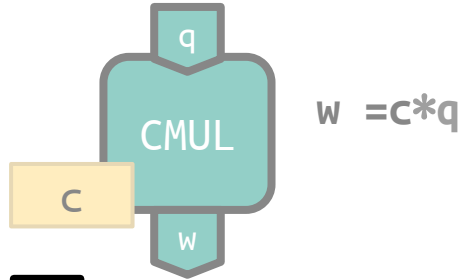
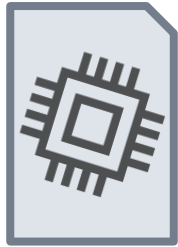
Assembly

Place
and Route



var v = $\int -0.25 * p \ dt_{ds}$ v(0)=0

observe p



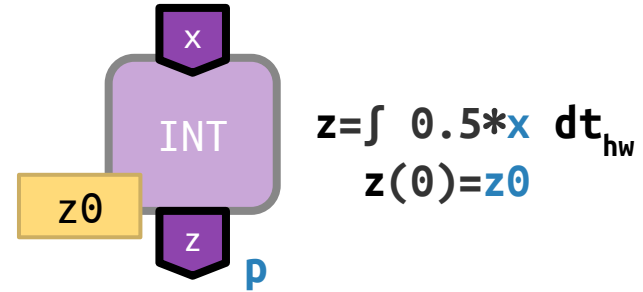
Circuit Synthesis

Subcircuit
Synthesis

Assembly

Place
and Route

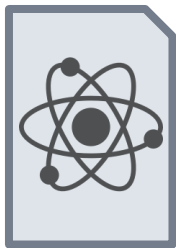
Unification



var p = $\int v \ dt_{ds}$ p(0)=10

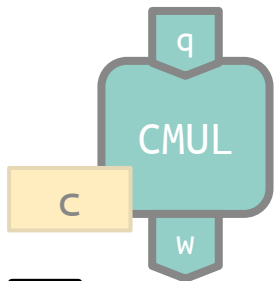
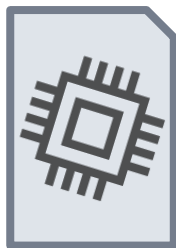
Goal

z = p

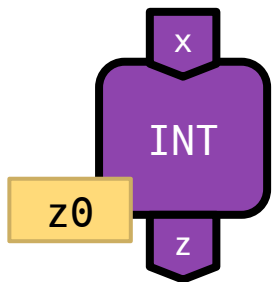


var $v = \int -0.25 * p \ dt_{ds} \quad v(0)=0$

observe p

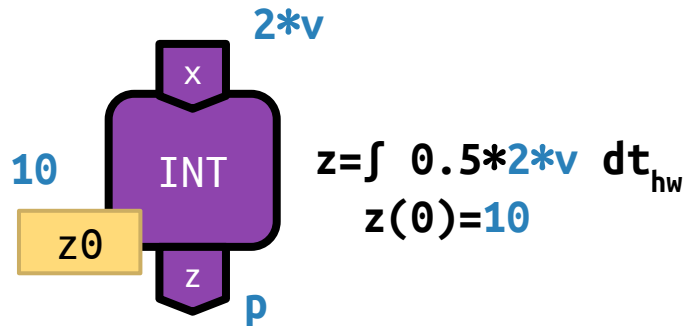


$w = c * q$



$z = \int 0.5 * x \ dt_{hw}$
 $z(0) = z0$

Unification



$z = \int 0.5 * 2 * v \ dt_{hw}$
 $z(0) = 10$

var $p = \int v \ dt_{ds} \quad p(0) = 10$

Goal

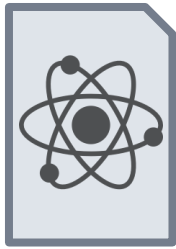
$z = p$

Circuit Synthesis

Subcircuit
Synthesis

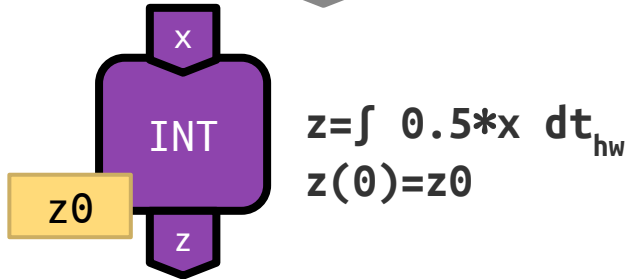
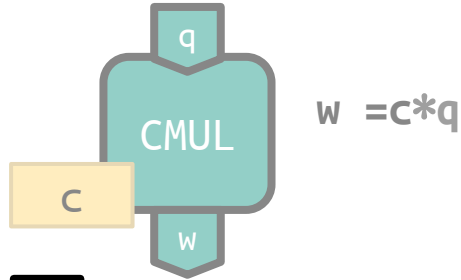
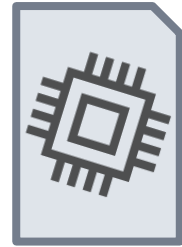
Assembly

Place
and Route



var v = $\int -0.25 * p \ dt_{ds}$ $v(0)=0$

observe p



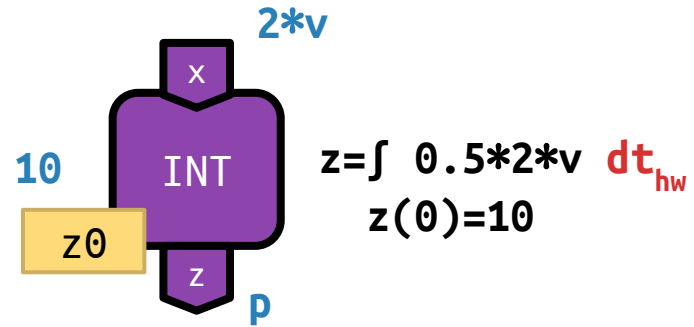
Circuit Synthesis

Subcircuit
Synthesis

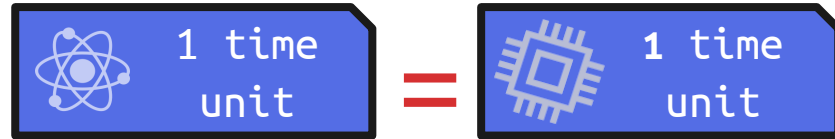
Assembly

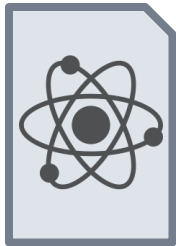
Place
and Route

Unification



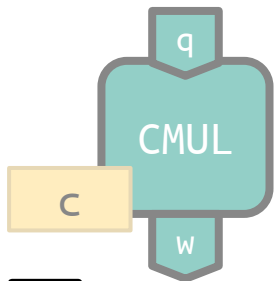
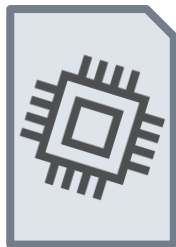
var p = $\int v \ dt_{ds}$ $p(0)=10$



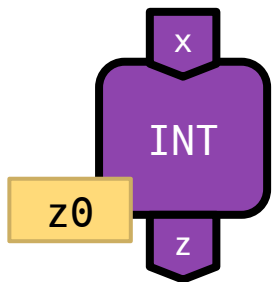


var $v = \int -0.25 * p \ dt_{ds} \quad v(0)=0$

observe p

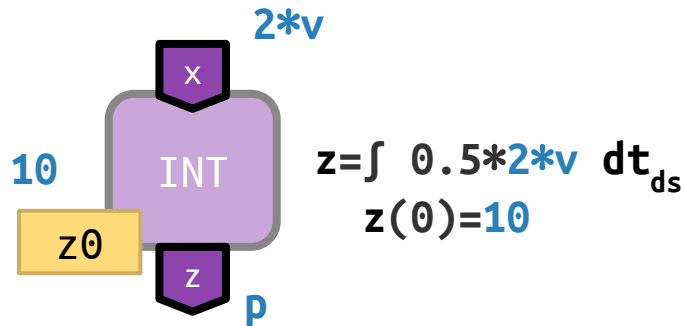


$w = c * q$



$z = \int 0.5 * x \ dt_{hw}$
 $z(0) = z0$

Unification



var $p = \int v \ dt_{ds} \quad p(0)=10$

Goal

$z = p$

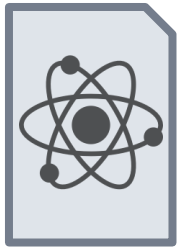


Circuit Synthesis

Subcircuit
Synthesis

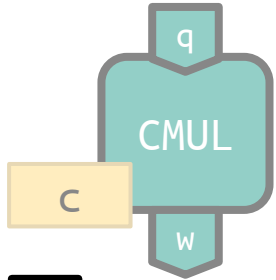
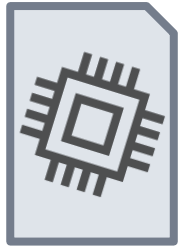
Assembly

Place
and Route

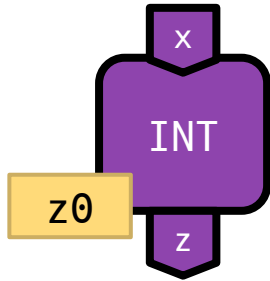


var $v = \int -0.25 * p \, dt_{ds} \quad v(0)=0$

observe p

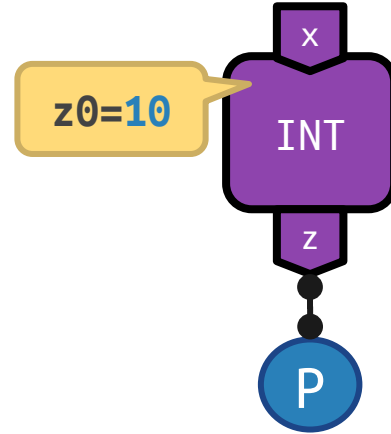


$$w = c * q$$



$$z = \int 0.5 * x \, dt_{hw}$$
$$z(0) = z_0$$

$$x = 2 * v$$

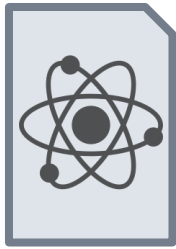


Circuit Synthesis

Subcircuit
Synthesis

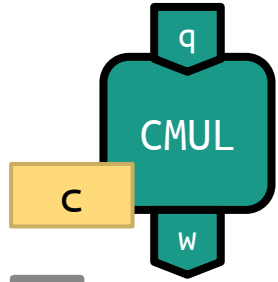
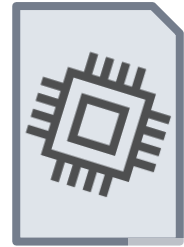
Assembly

Place
and Route

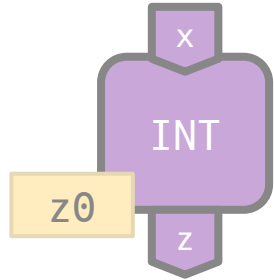


var $v = \int -0.25 * p \ dt_{ds} \quad v(0)=0$

observe p



$$w = c * q$$

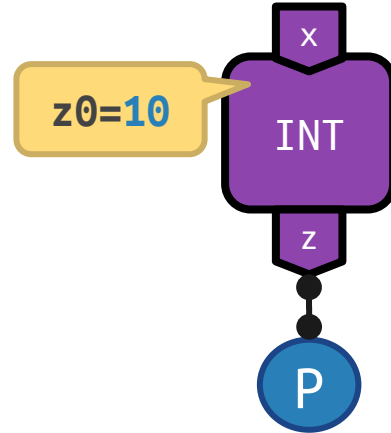


$$z = \int 0.5 * x \ dt_{hw}$$

$$z(0) = z0$$

Block Selection

$$x = 2 * v$$

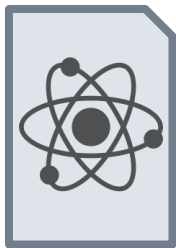


Circuit Synthesis

Subcircuit
Synthesis

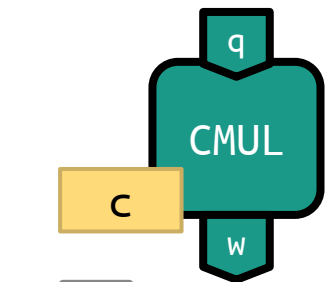
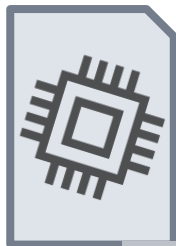
Assembly

Place
and Route

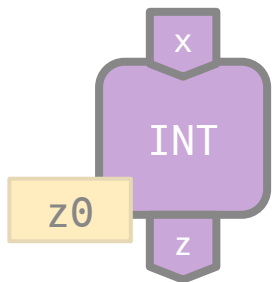


var $v = \int -0.25 * p \, dt_{ds} \quad v(0)=0$

observe p

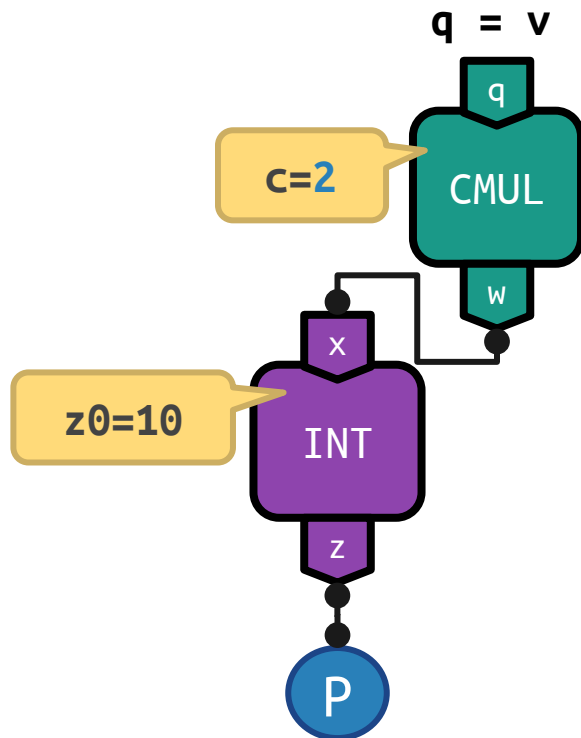


$w = c * q$



$z = \int 0.5 * x \, dt_{hw}$
 $z(0) = z0$

Unification

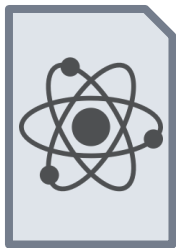


Circuit Synthesis

Subcircuit
Synthesis

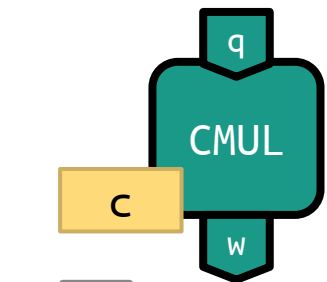
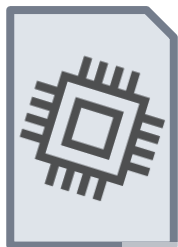
Assembly

Place
and Route

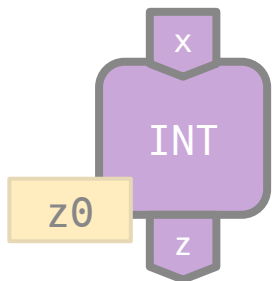


var $v = \int -0.25 * p \ dt_{ds} \quad v(0)=0$

observe p



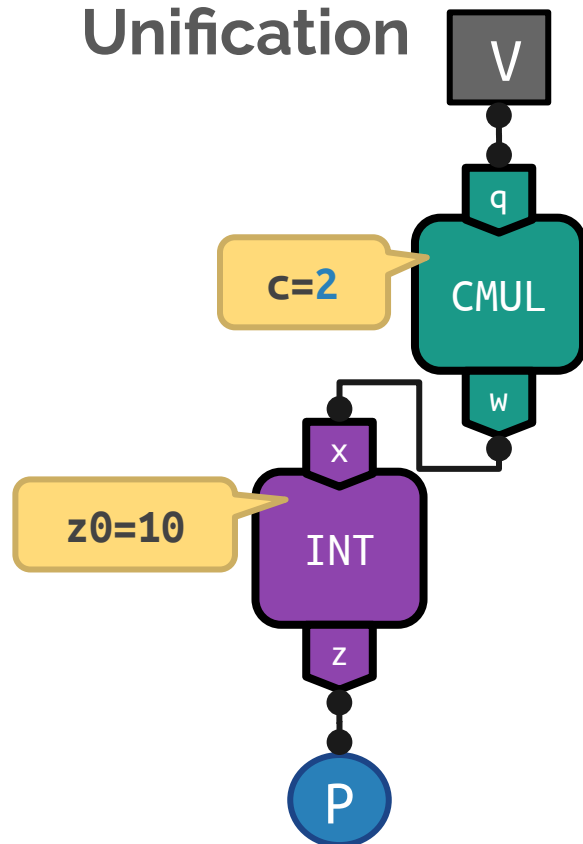
$$w = c * q$$



$$z = \int 0.5 * x \ dt_{hw}$$

$$z(0) = z_0$$

Unification

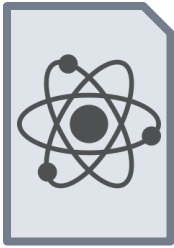


Circuit Synthesis

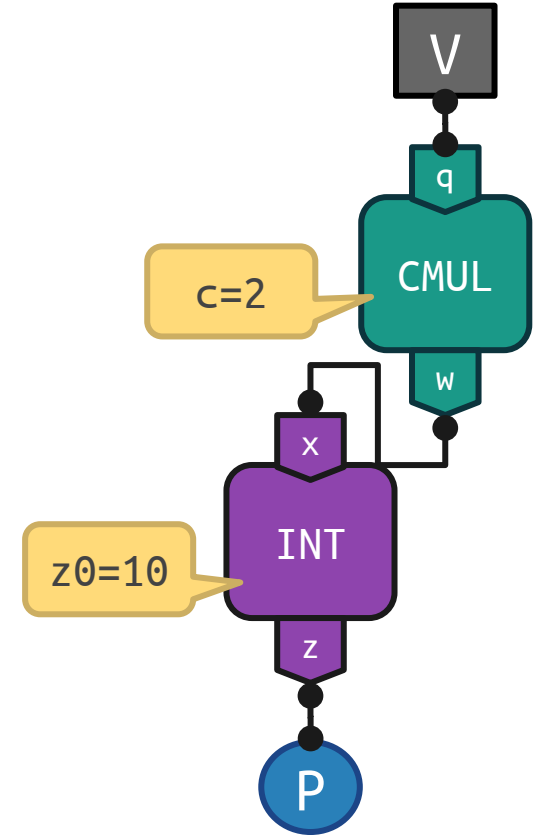
Subcircuit
Synthesis

Assembly

Place
and Route



P $p = \int 0.5 * (2 * v) dt_{ds}$
 $p(0) = 10$

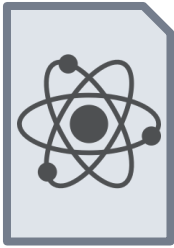


Circuit Synthesis


Subcircuit
Synthesis

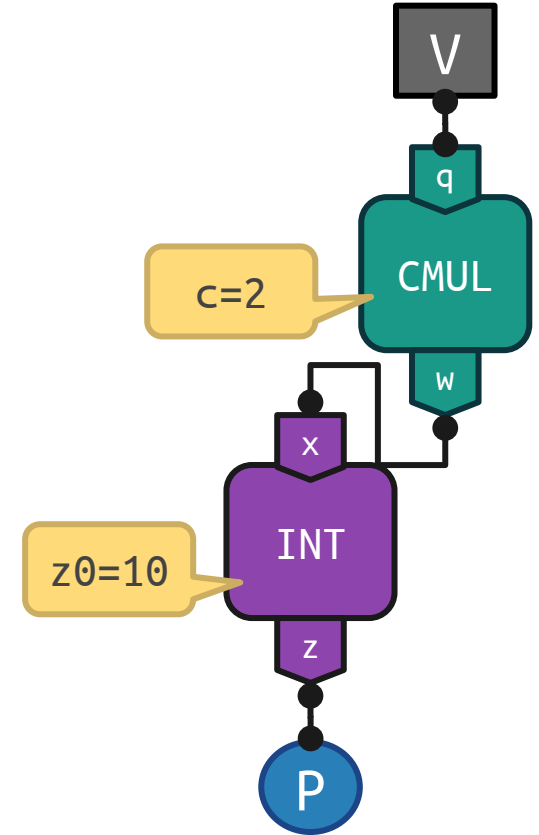
Assembly

Place
and Route



P $p = \int 0.5 * (2 * v) dt_{ds}$
 $p(0) = 10$

 $p = \int v dt_{ds}$
 $p(0) = 10$

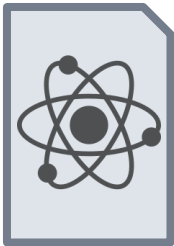


Circuit Synthesis

Subcircuit
Synthesis

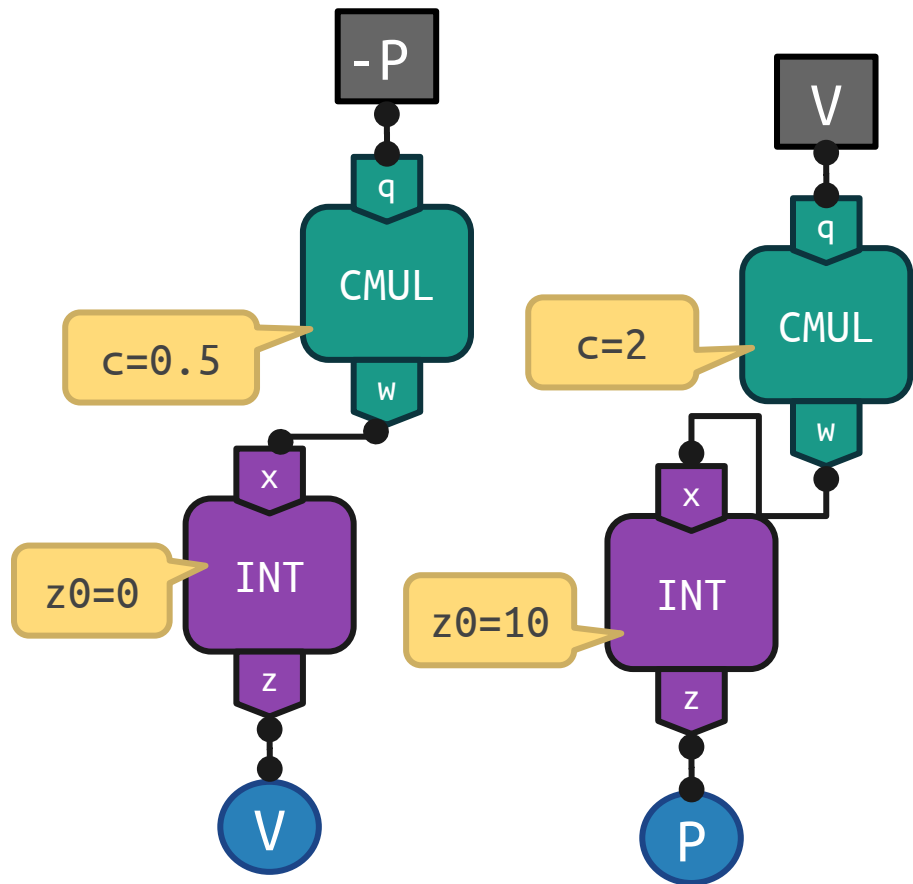
Assembly

Place
and Route



V

$$v = \int 0.5 * (0.5 * (-p)) dt_{ds}$$
$$v(0) = 0$$

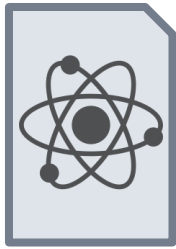


Circuit Synthesis

Subcircuit
Synthesis

Assembly

Place
and Route

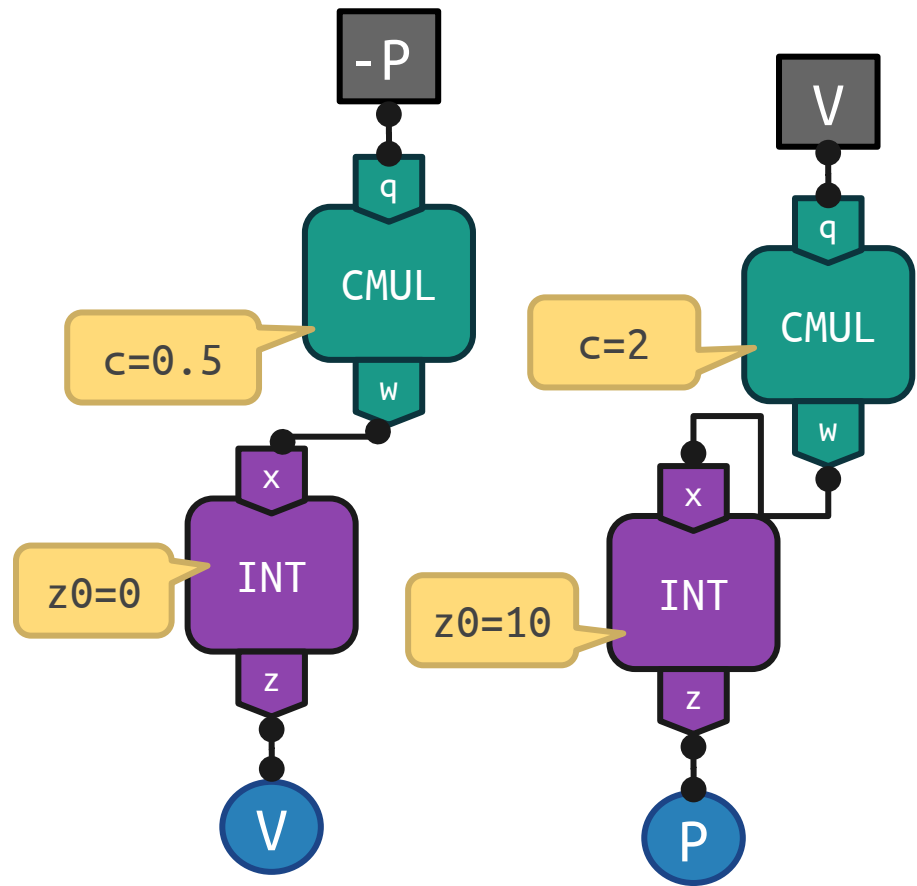


V

$$v = \int 0.5 * (0.5 * (-p)) dt_{ds}$$
$$v(0) = 0$$



$$v = \int -0.25 * p dt_{ds}$$
$$v(0) = 0$$

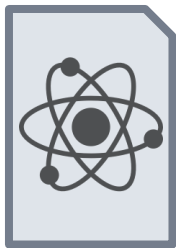


Circuit Synthesis

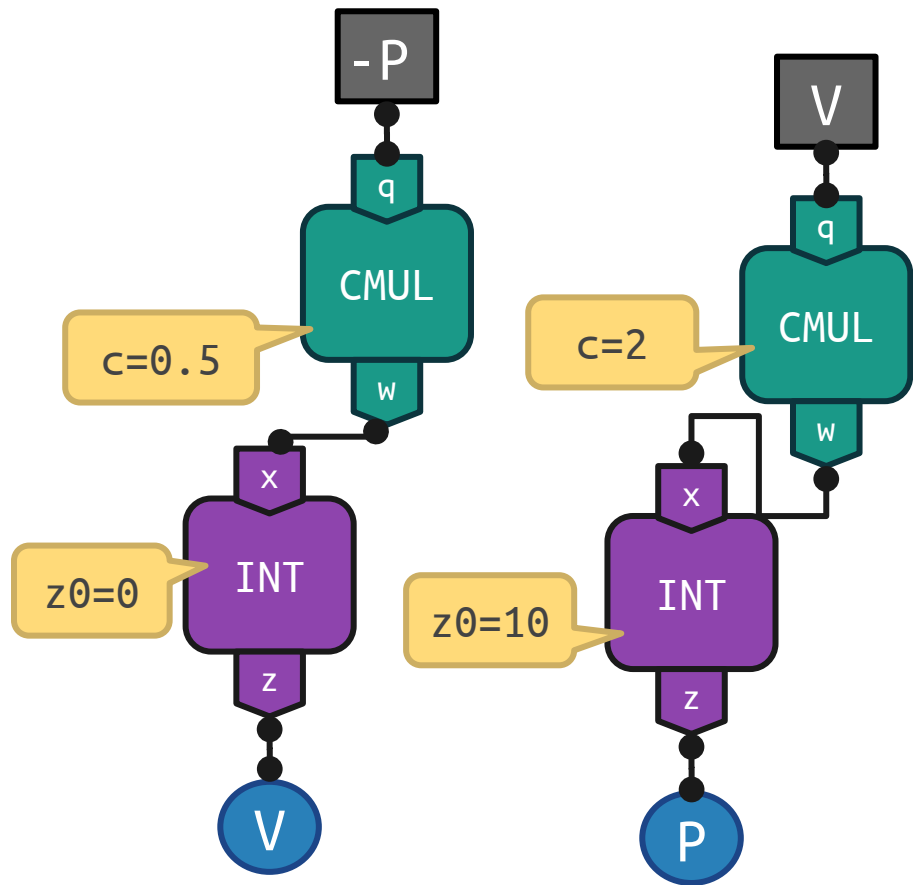
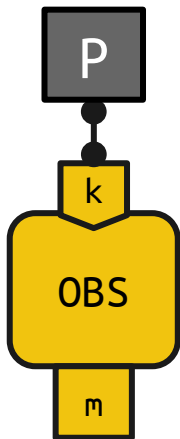
Subcircuit
Synthesis

Assembly

Place
and Route



observe p

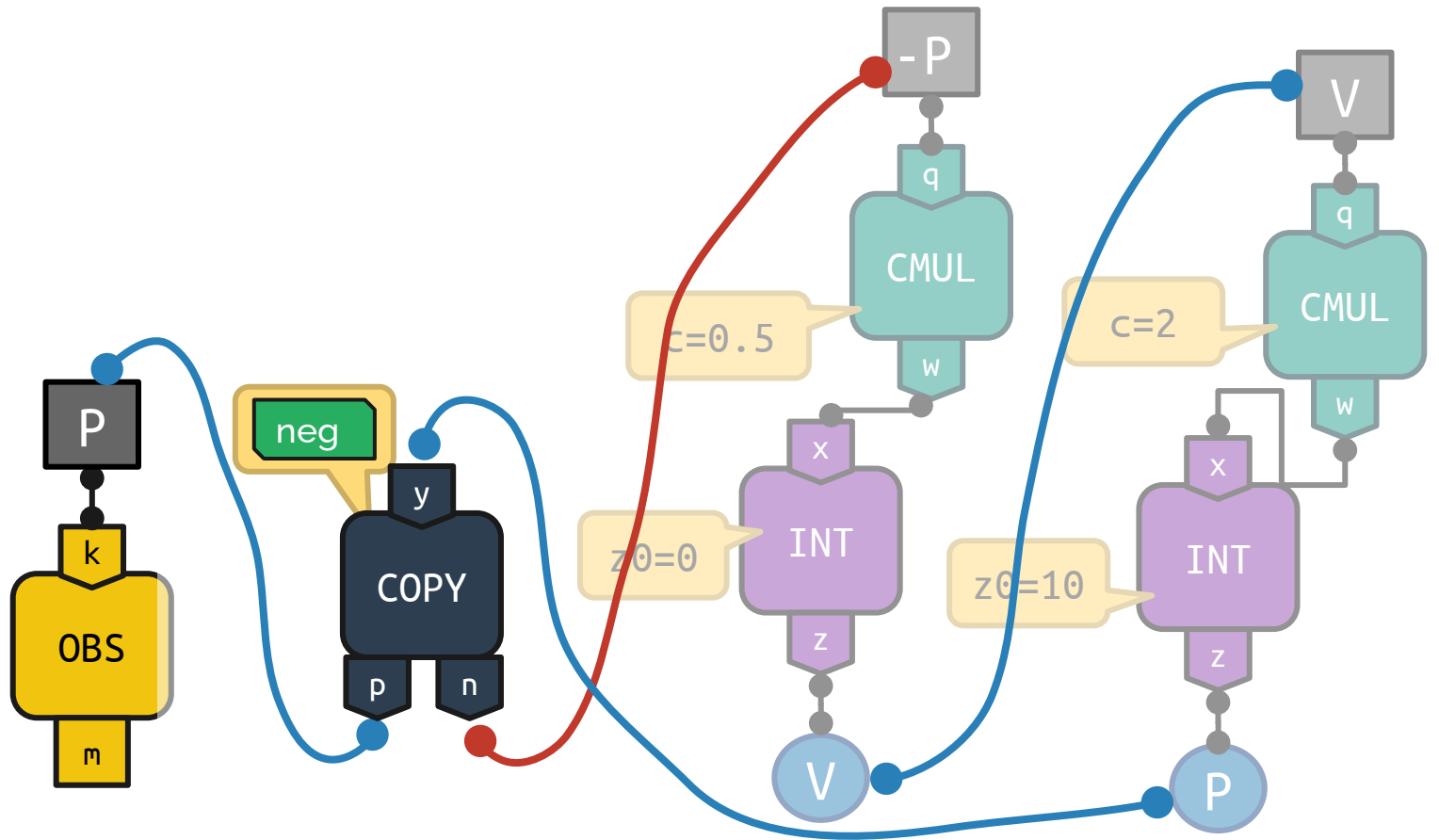


Circuit Synthesis

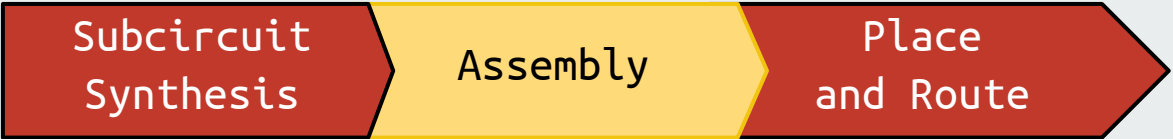
Subcircuit
Synthesis

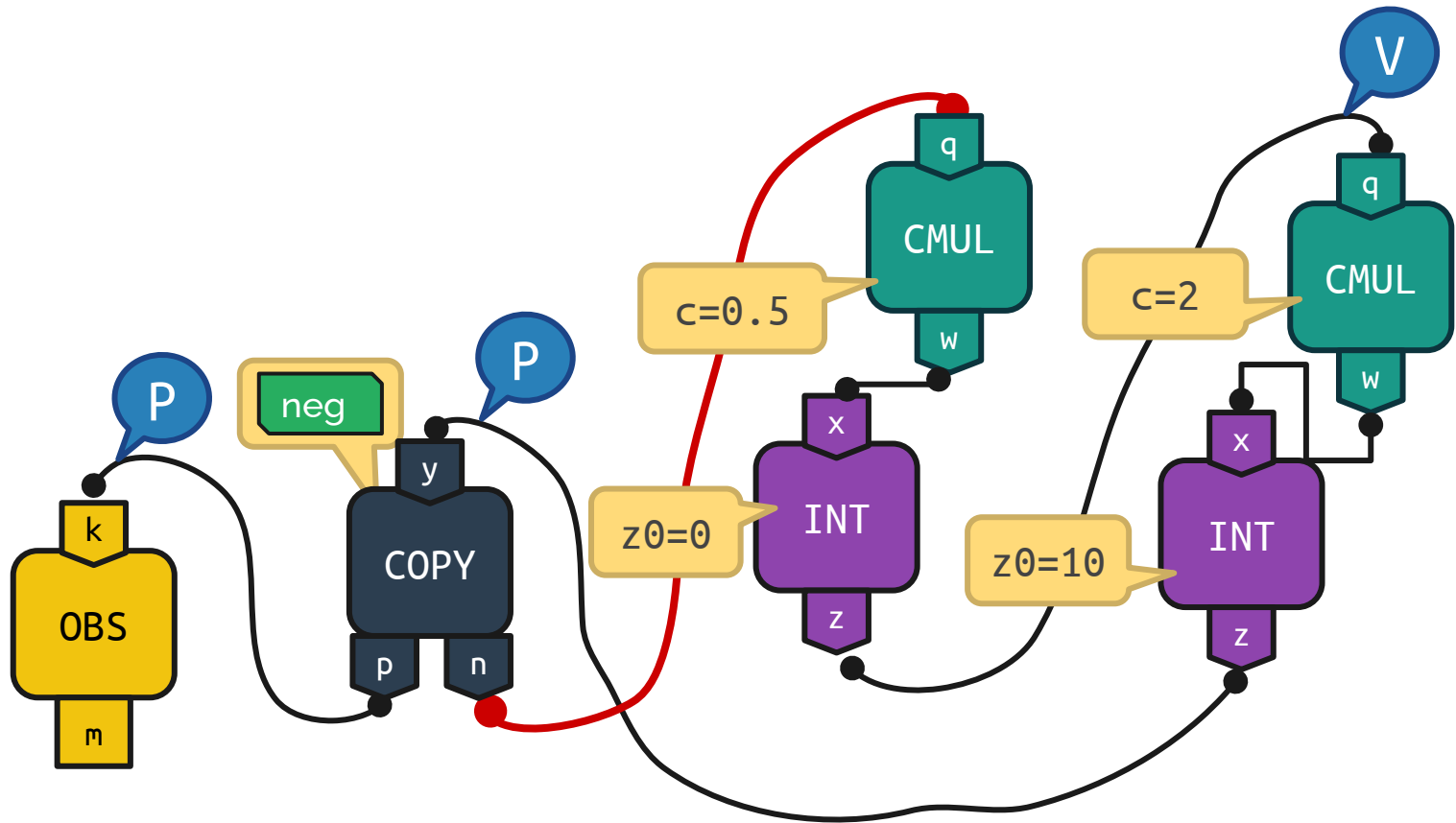
Assembly

Place
and Route

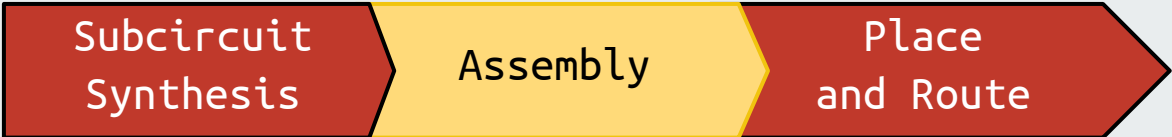


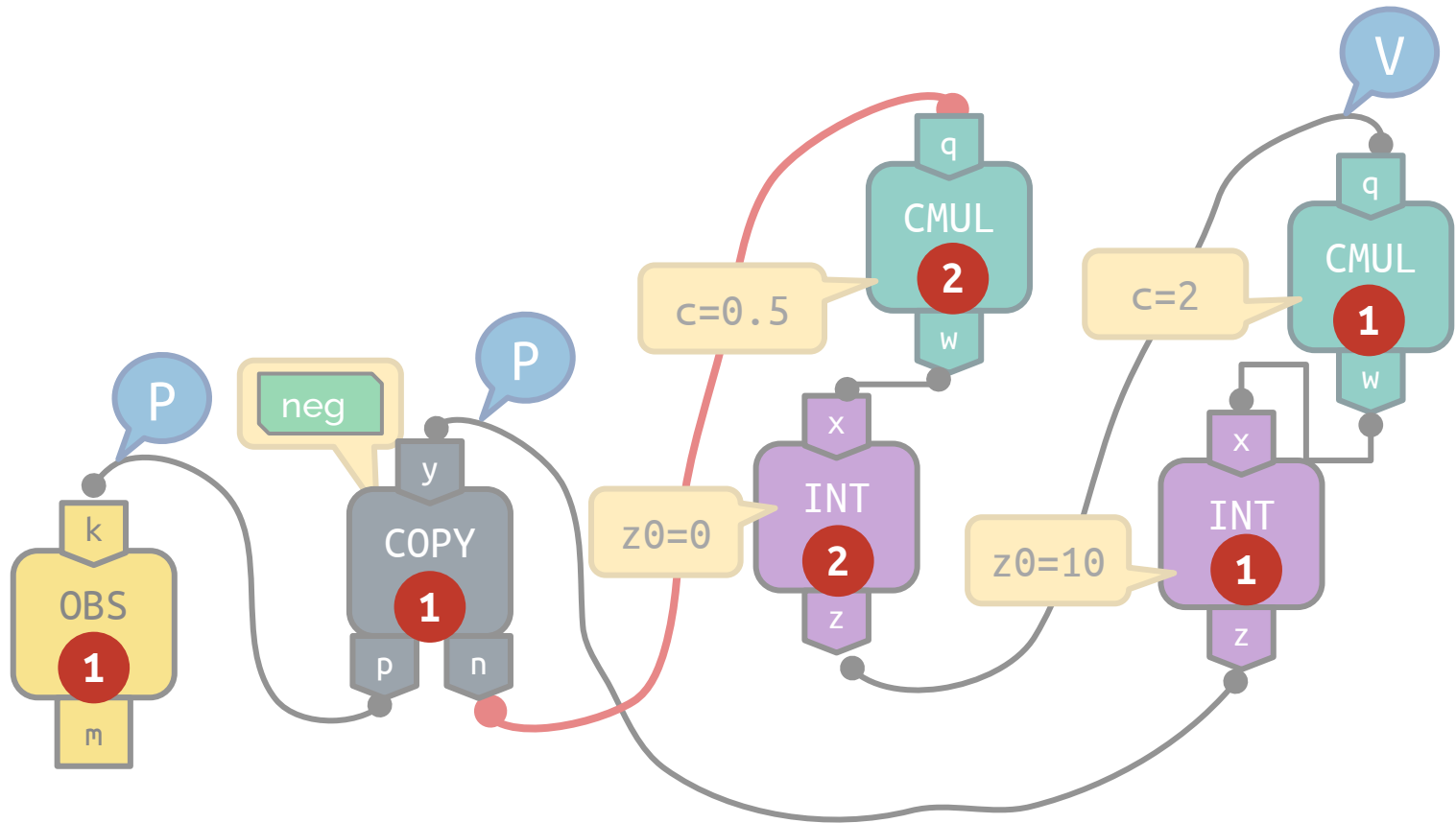
Circuit Synthesis





Circuit Synthesis



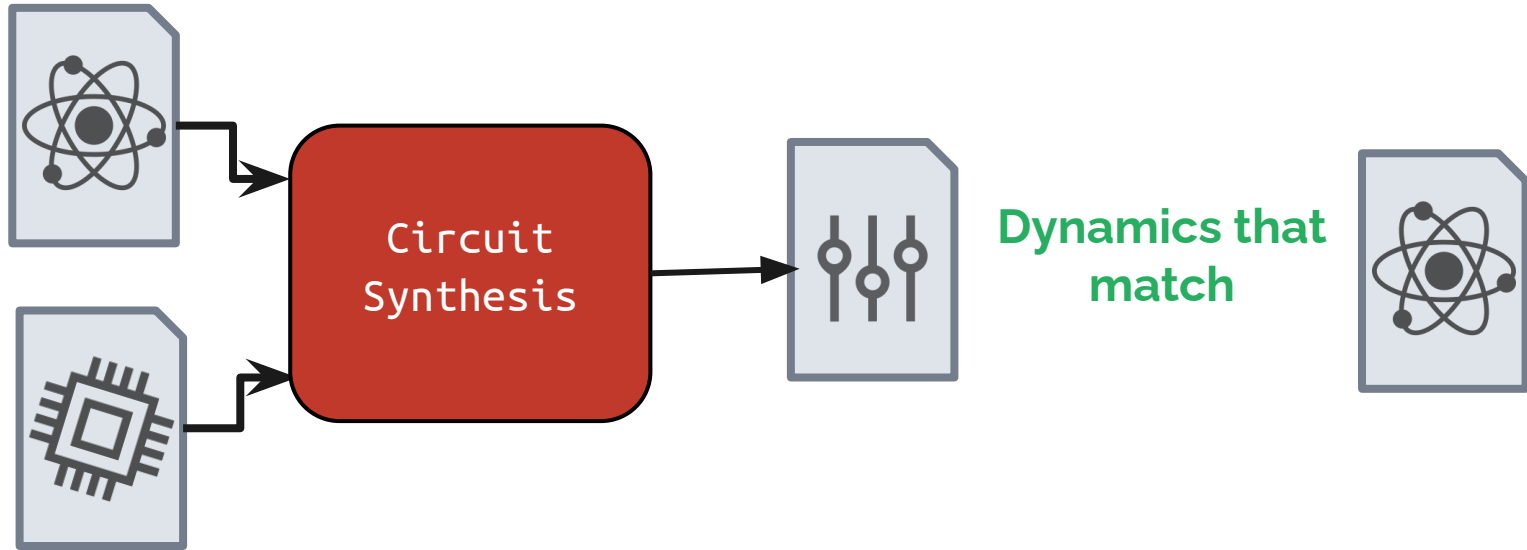


Circuit Synthesis

Subcircuit
Synthesis

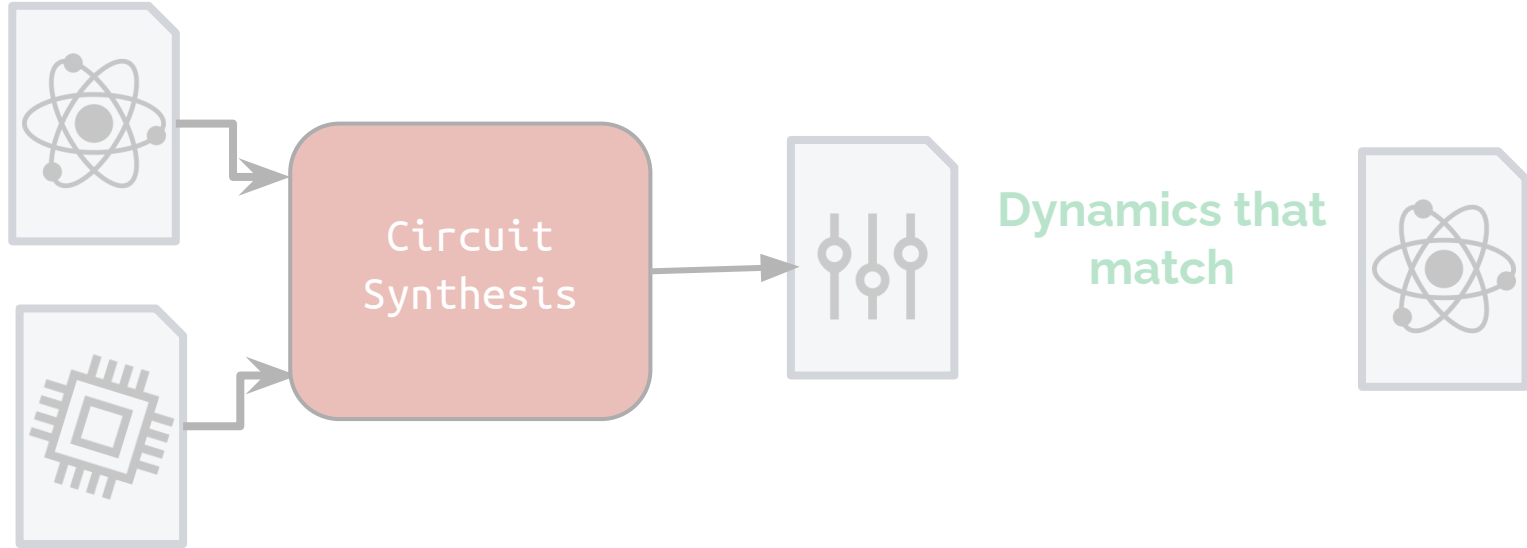
Assembly

Place
and Route

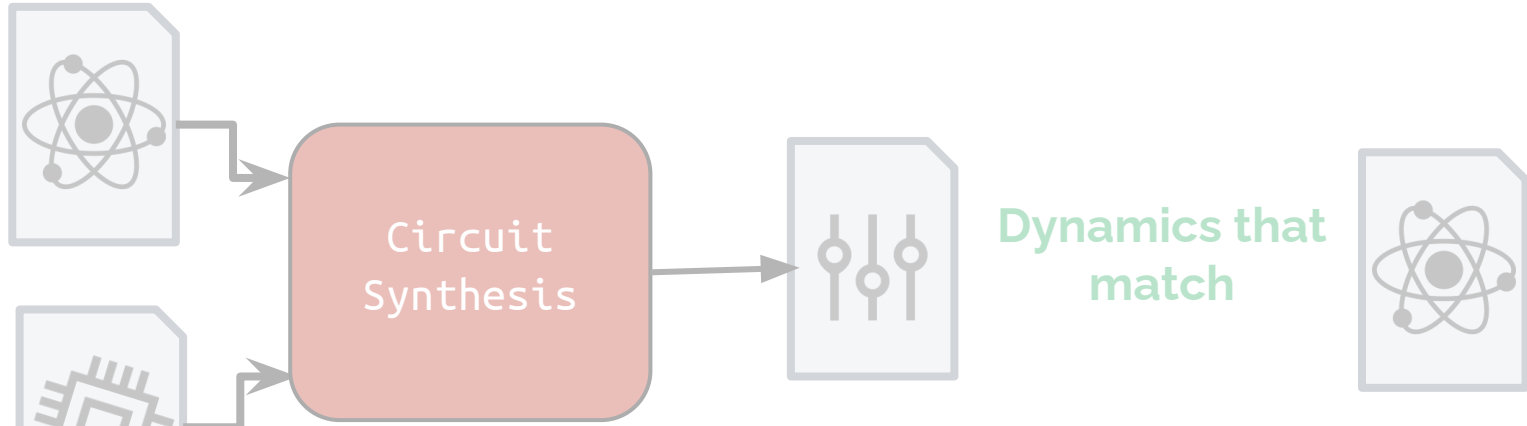


1. Configuration Synthesis for Programmable Analog Devices with Arco. PLDI 2016.
2. Noise-Aware Dynamical System Compilation for Analog Devices with Legno. ASPLOS 2020.

Are we done?



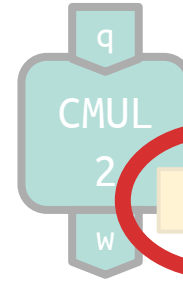
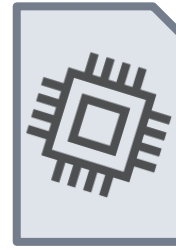
Are we done?



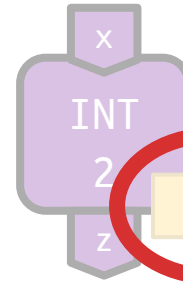
No! Must consider *analog behavior* of blocks

Physical Limitations

Operating range Limitations



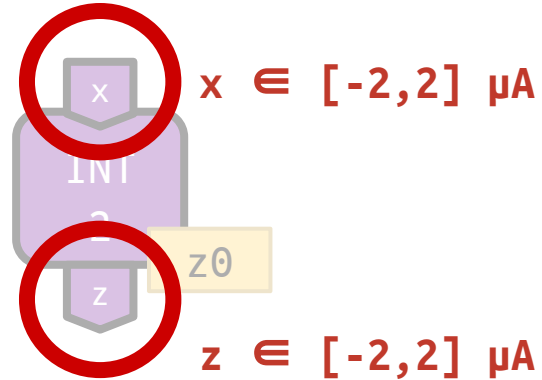
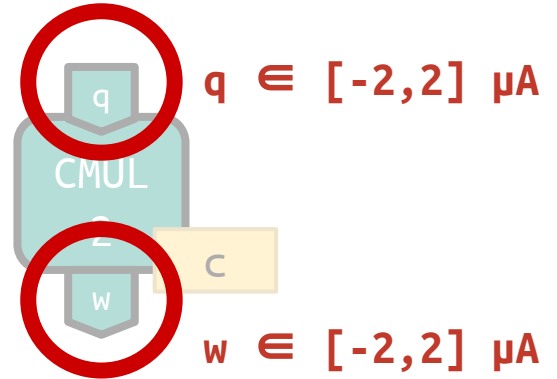
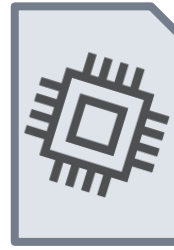
$$c \in [-1,1]$$



$$z_0 \in [-1,1]$$

Physical Limitations

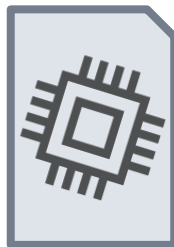
Operating range Limitations



Inputs



Execution Speed Limitations



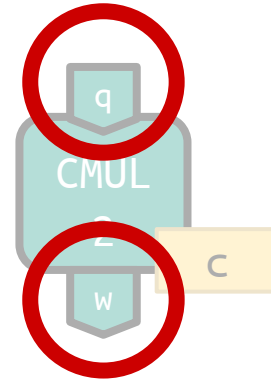
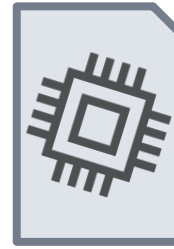
=

at least

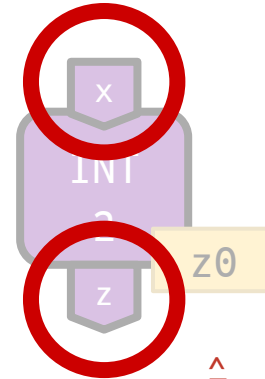


Physical Limitations

Analog Noise



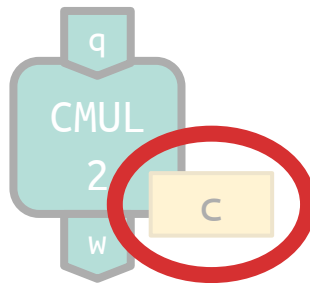
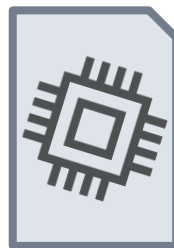
$$\hat{w} \sim \text{Normal}(\bar{w}, 0.02 \mu\text{A})$$



$$\hat{z} \sim \text{Normal}(\bar{z}, 0.01 \mu\text{A})$$

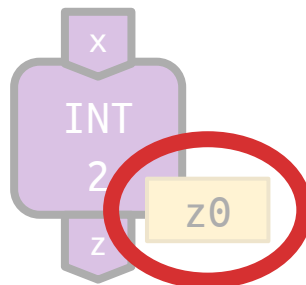
Physical Limitations

Quantization Error



$$c \in [-1, -0.99\dots, 0.99, 1]$$

$$\text{error} = 0.01$$



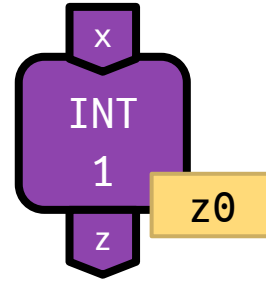
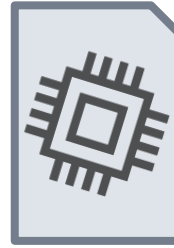
$$z_0 \in [-1, -0.99\dots, 0.99, 1]$$

$$\text{error} = 0.01$$

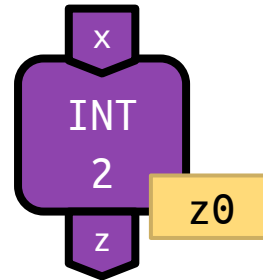
Inputs



Manufacturing Variations



$$z = \int 0.64 * 0.5 * x \, dt_{hw}$$
$$z(0) = 0.80 * z_0$$

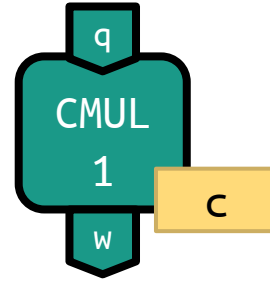
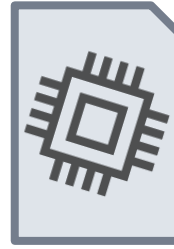


$$z = \int 0.93 * 0.5 * x \, dt_{hw}$$
$$z(0) = 0.77 * z_0$$

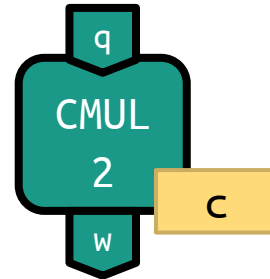
Inputs



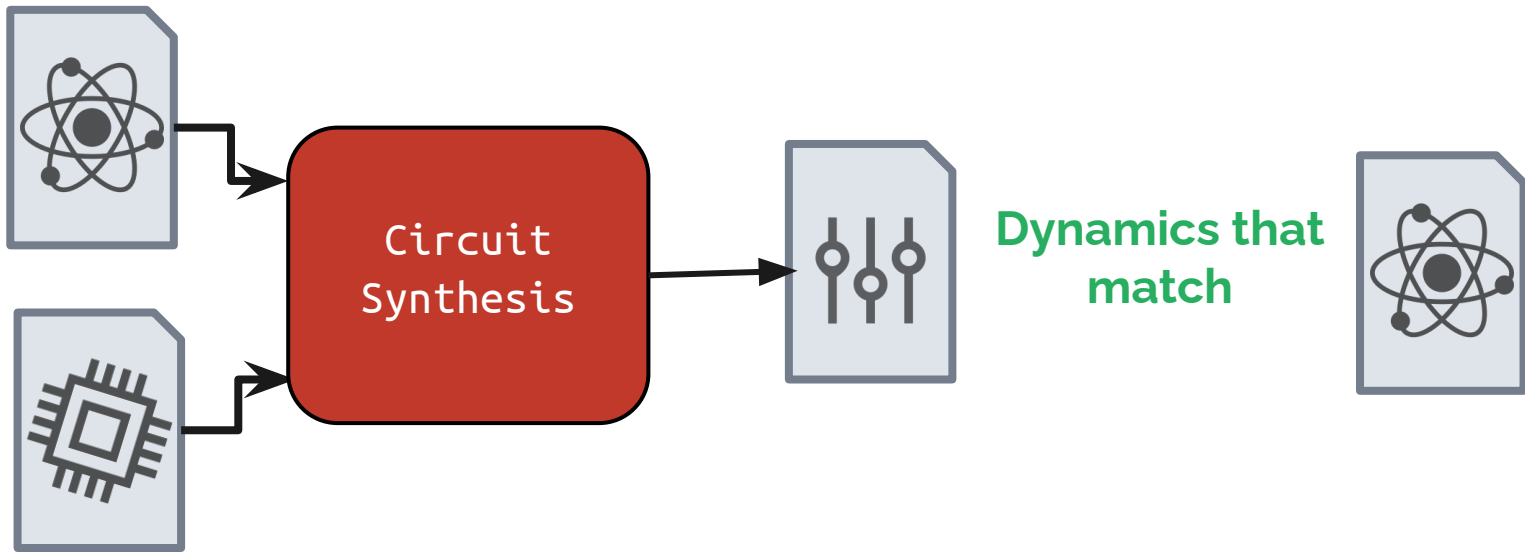
Manufacturing Variations

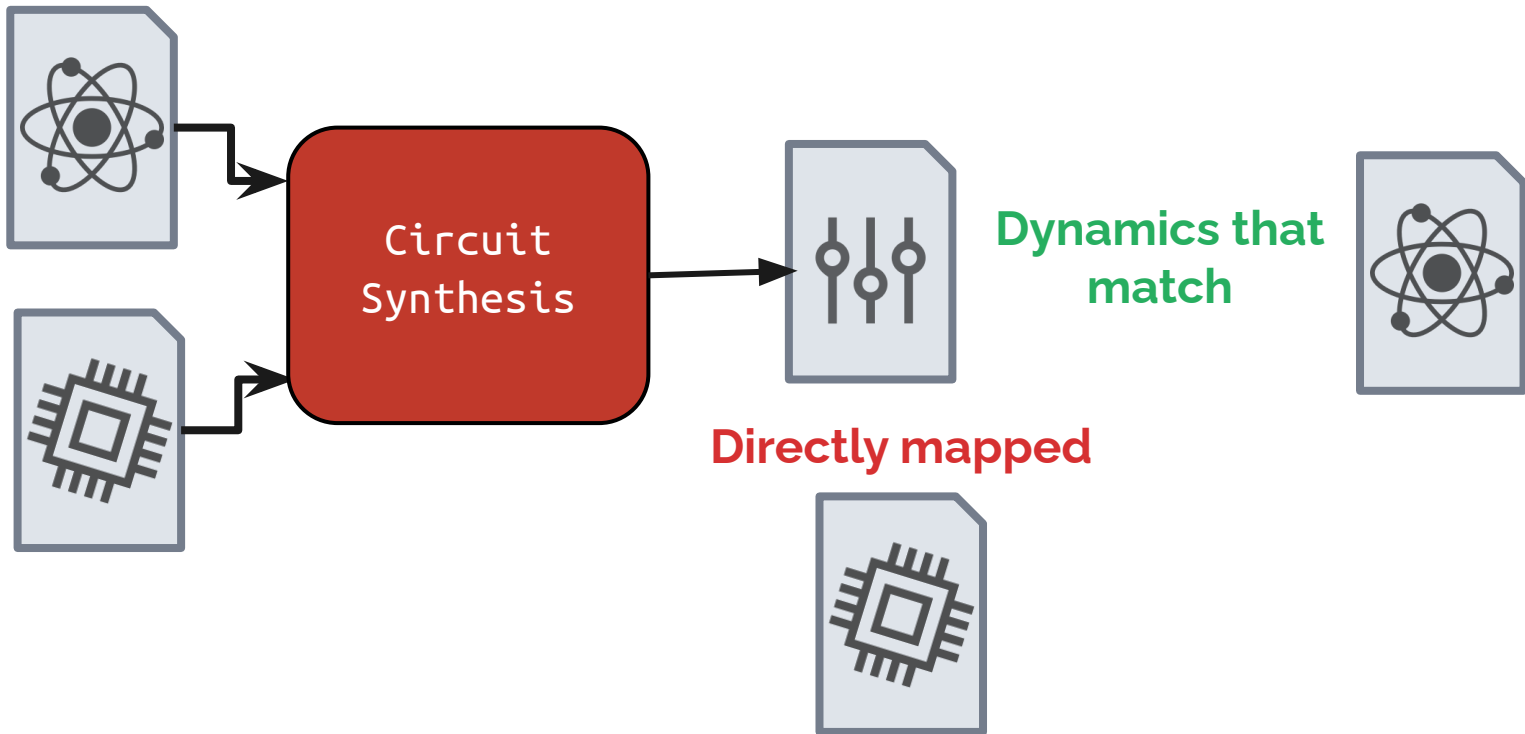


$$w = 0.76 * c * q$$



$$w = 1.12 * c * q$$

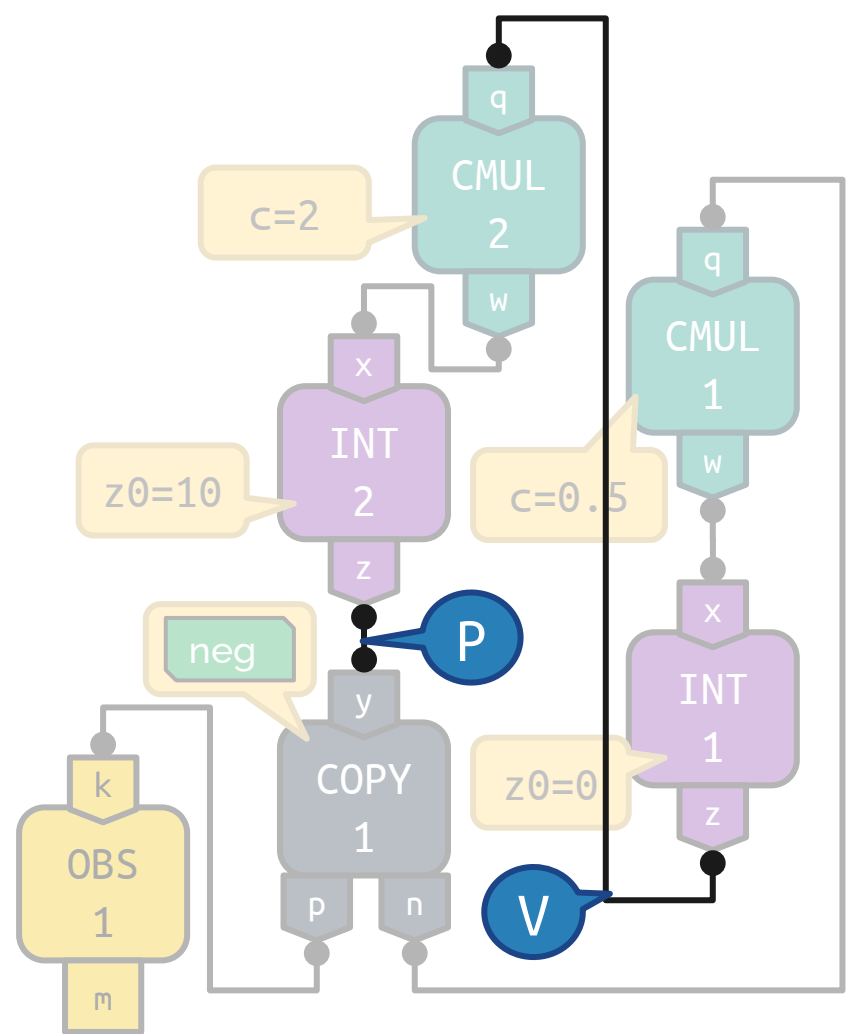




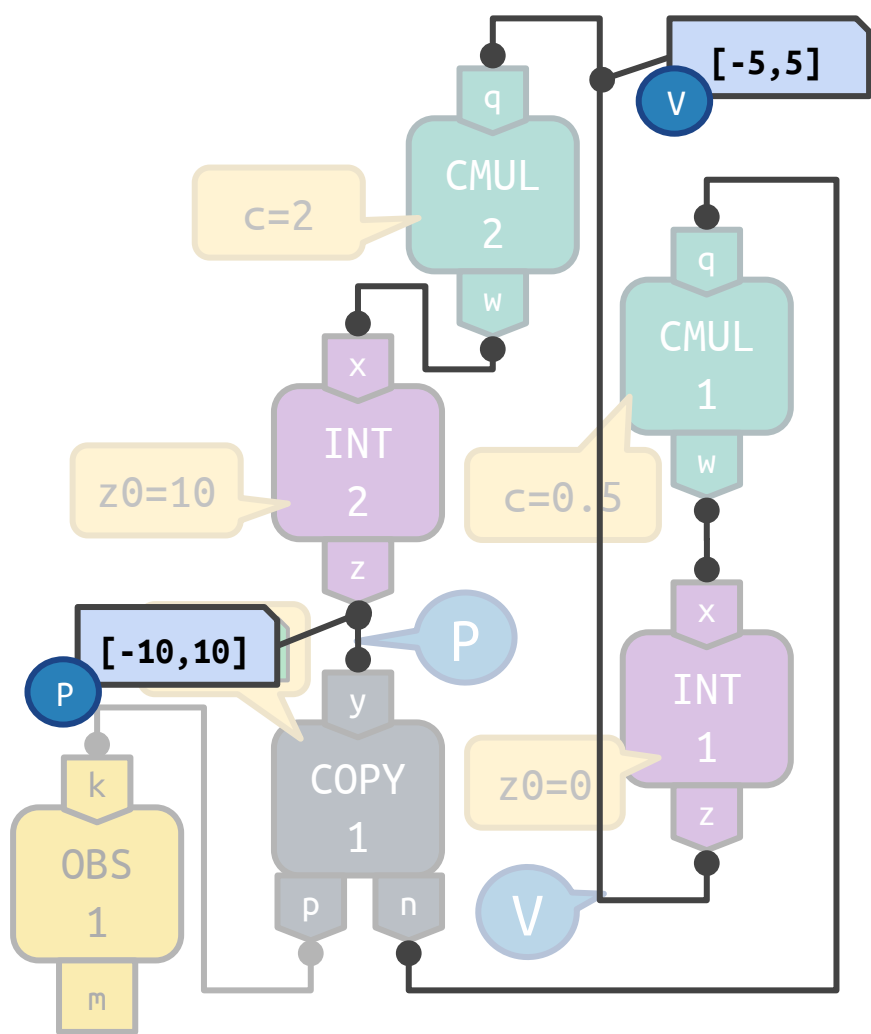
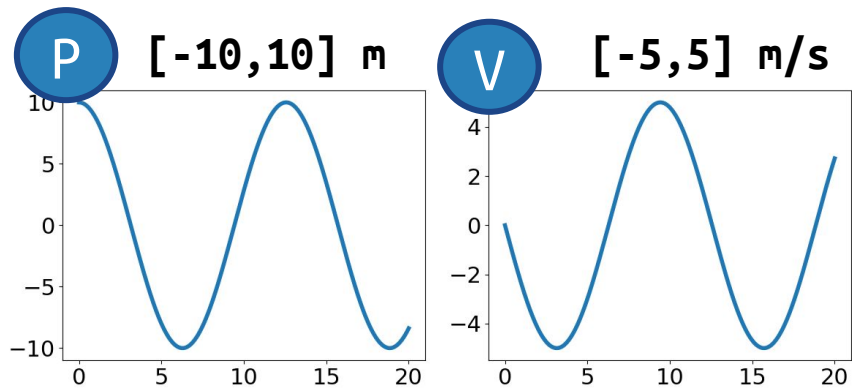
Directly Mapped Configuration

P 1 μA = 1 meter

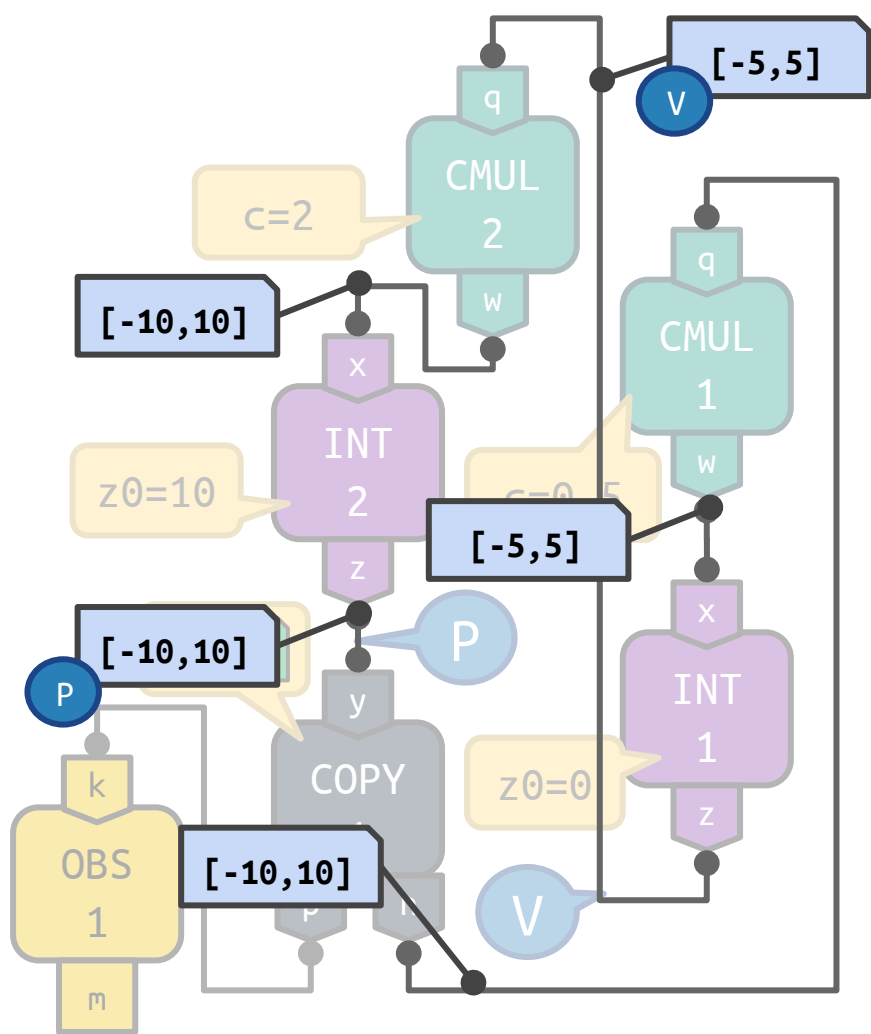
V 1 μA = 1 meter/second



Directly Mapped Configuration

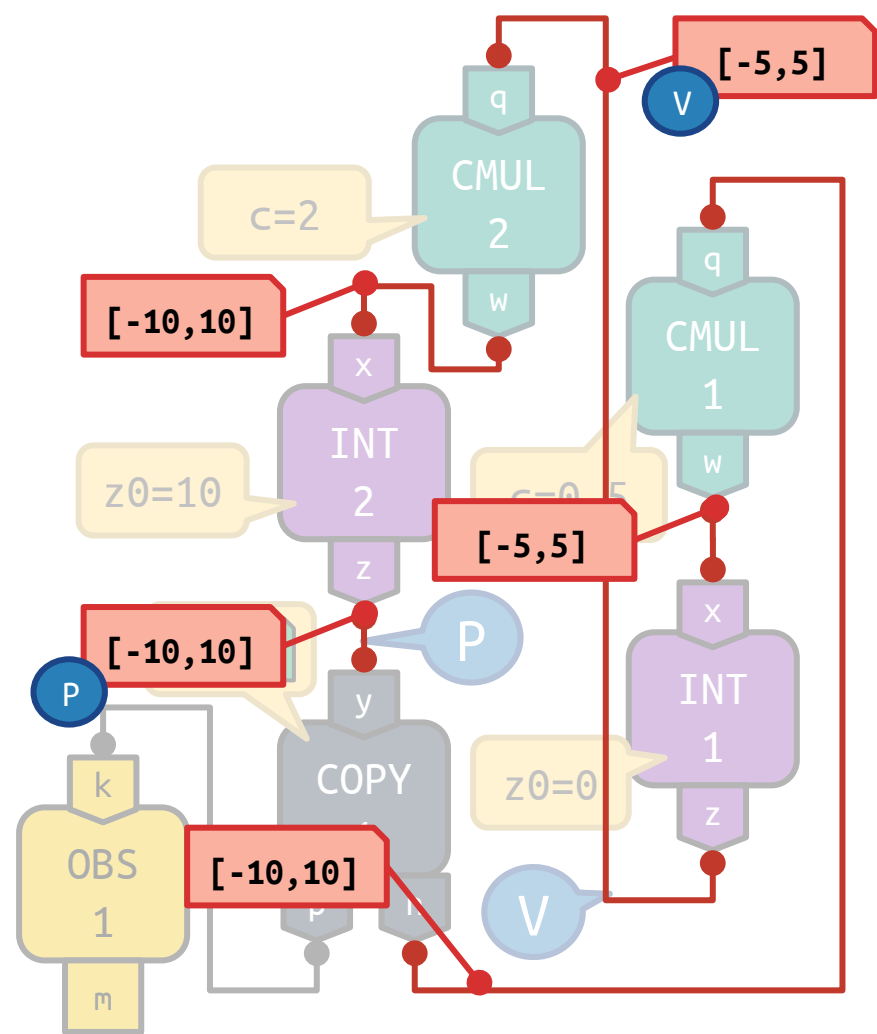


Directly Mapped Configuration



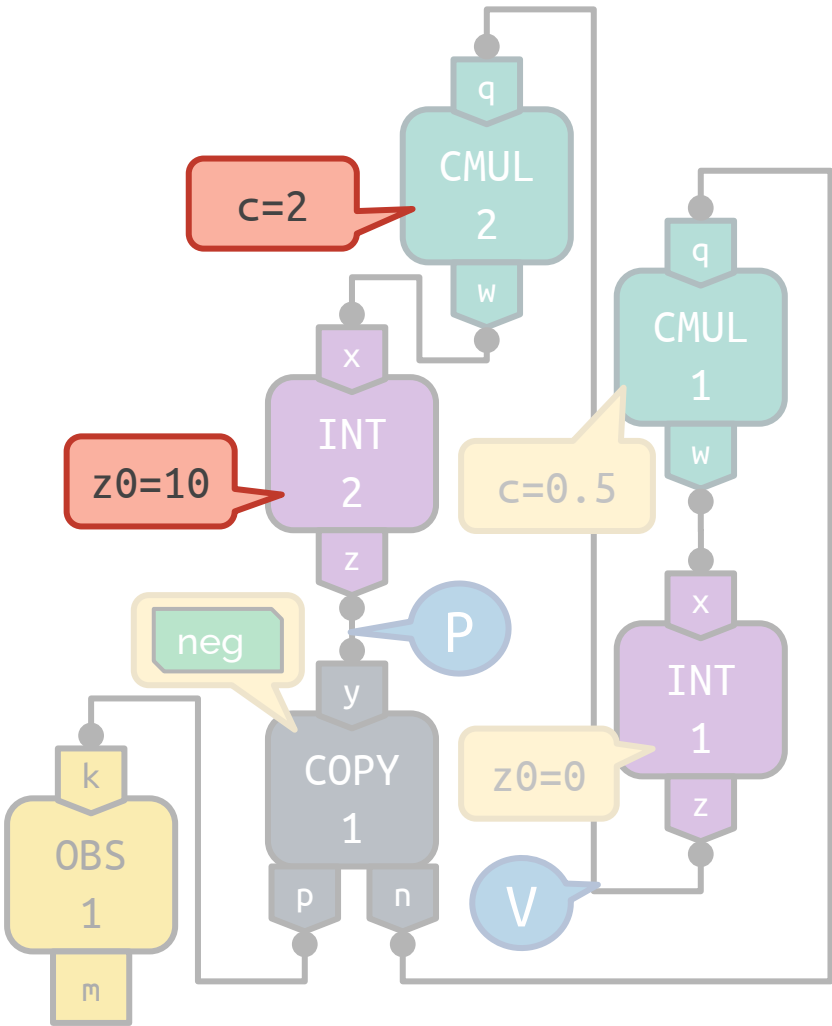
Operating Range Limitation Violations

$x, z, q, w \in [-2, 2] \mu\text{A}$

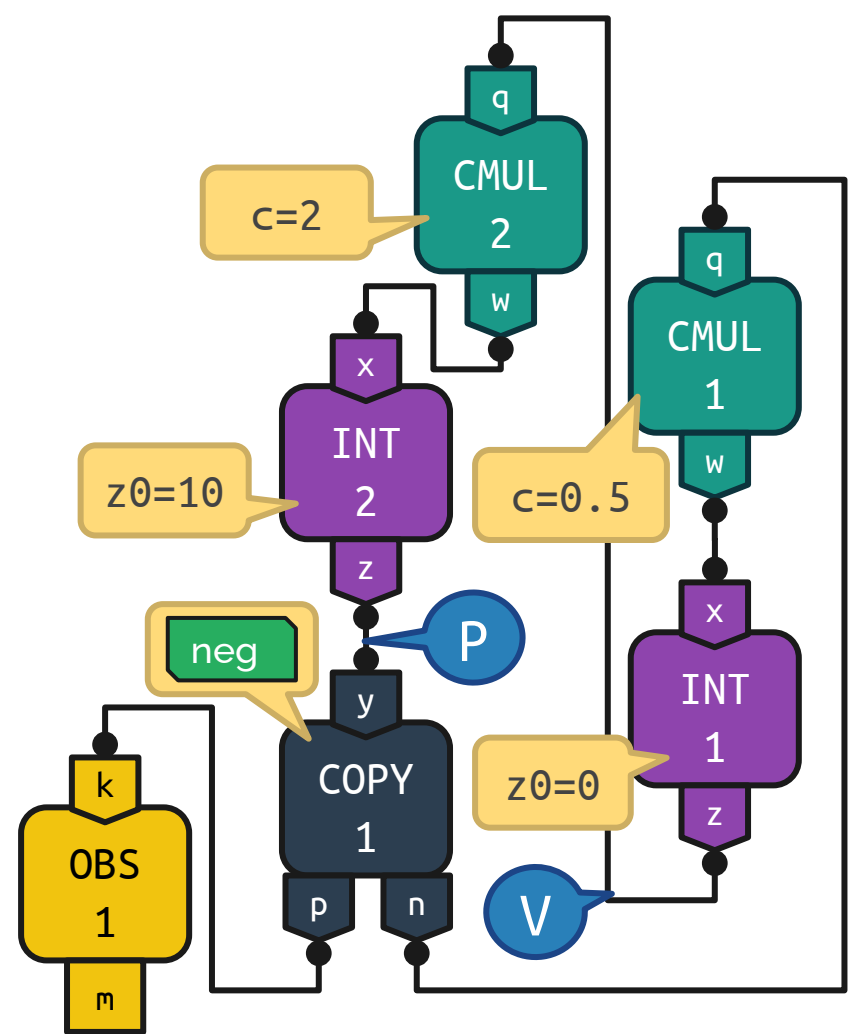
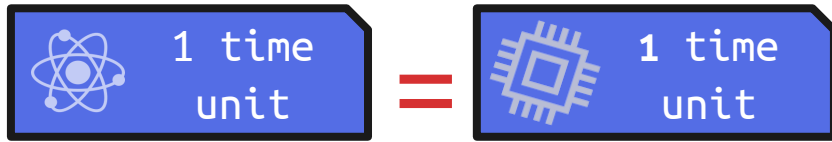


Operating Range Limitation Violations

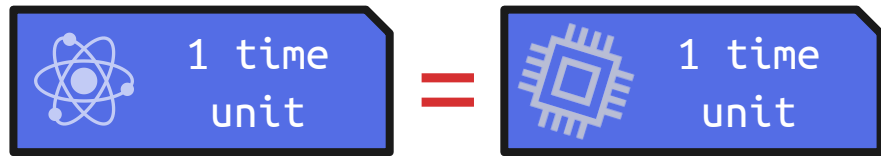
$$c, z_0 \in [-1, 1]$$



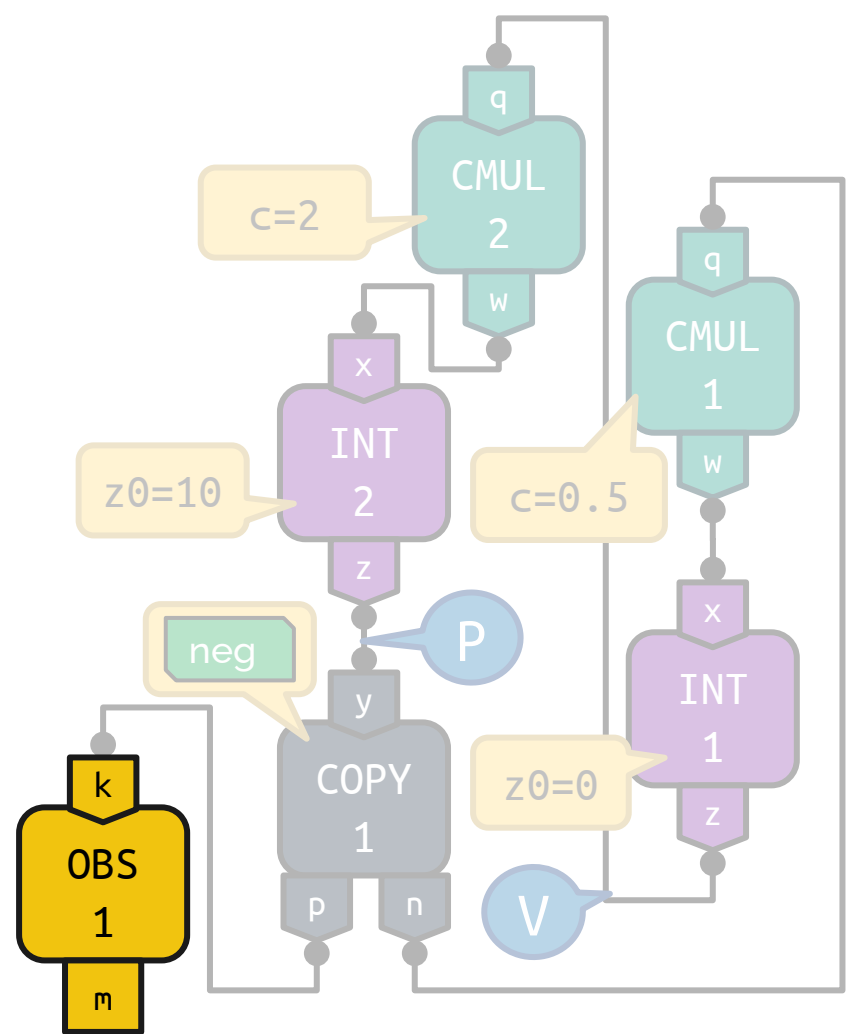
Directly Mapped Configuration



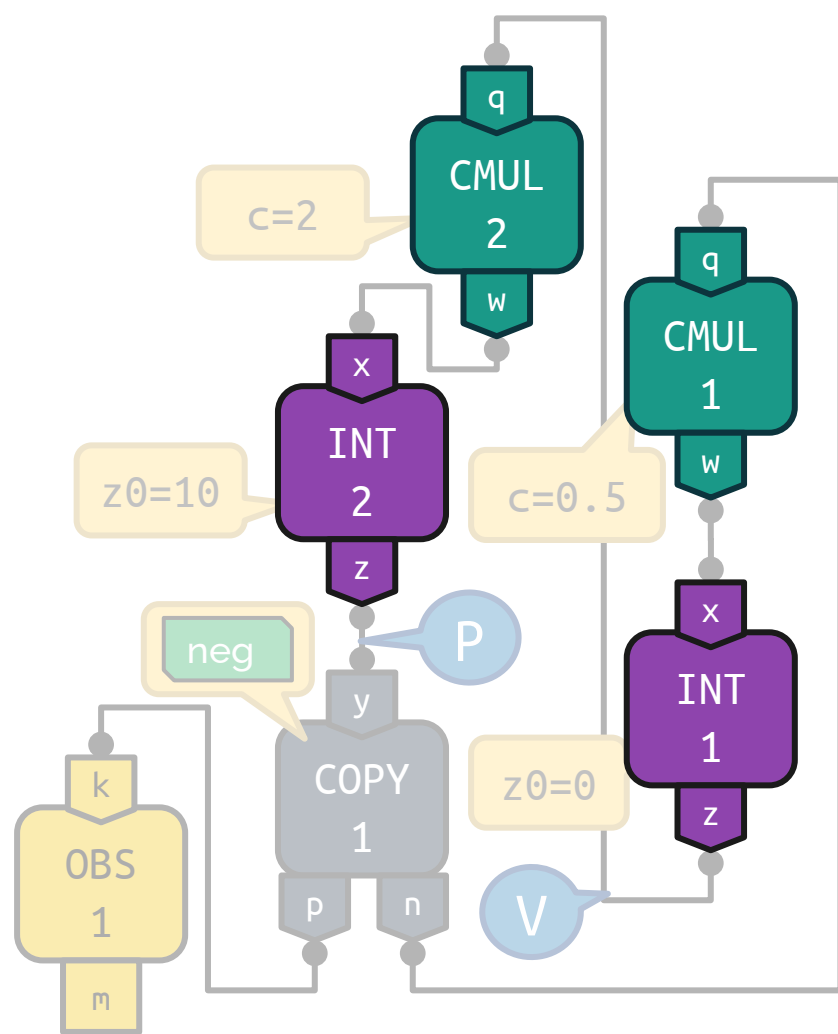
Execution Speed Limitations



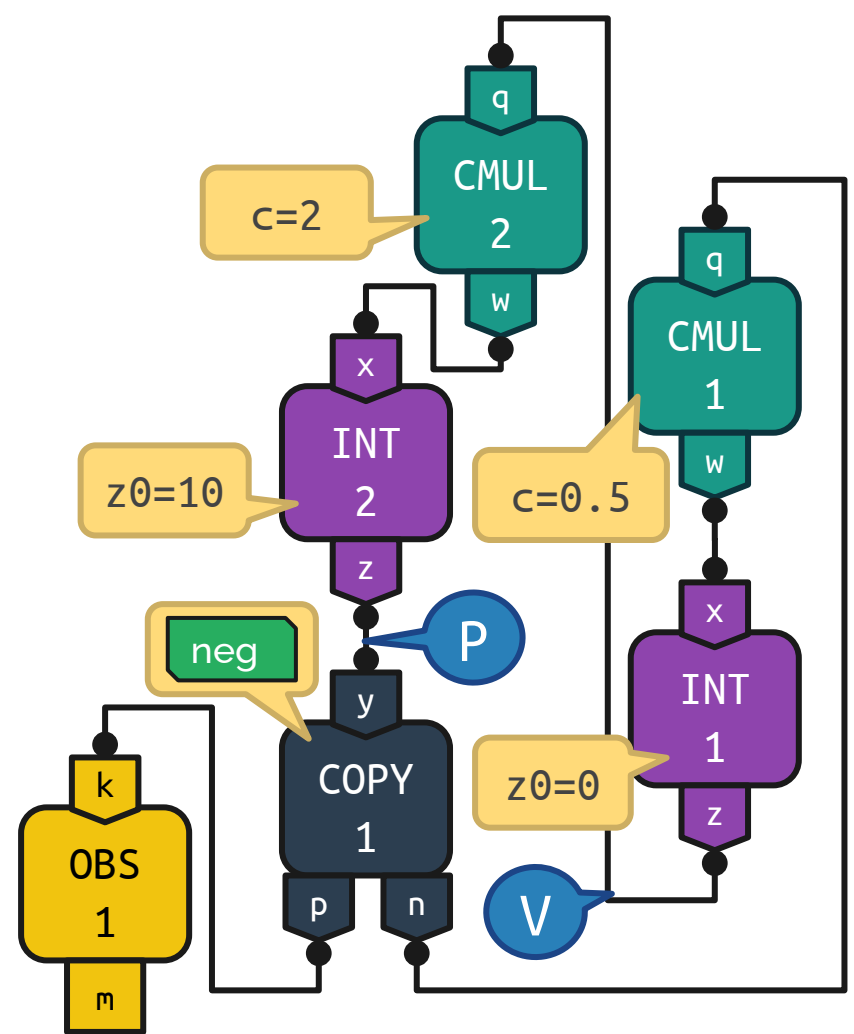
Must be at least



Manufacturing Variations



Noise and Quantization Error



How do we mitigate these behaviors?

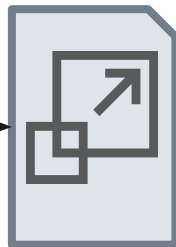
How do we mitigate these behaviors?

We Scale the Analog Device Configuration

Unscaled



Scaling Transform



Algebraically Equivalent to

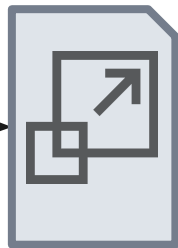


Automated Scaling

Unscaled



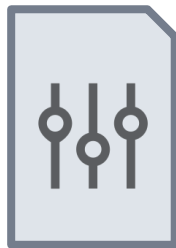
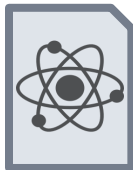
Scaling Transform



Scaled



Algebraically Equivalent to



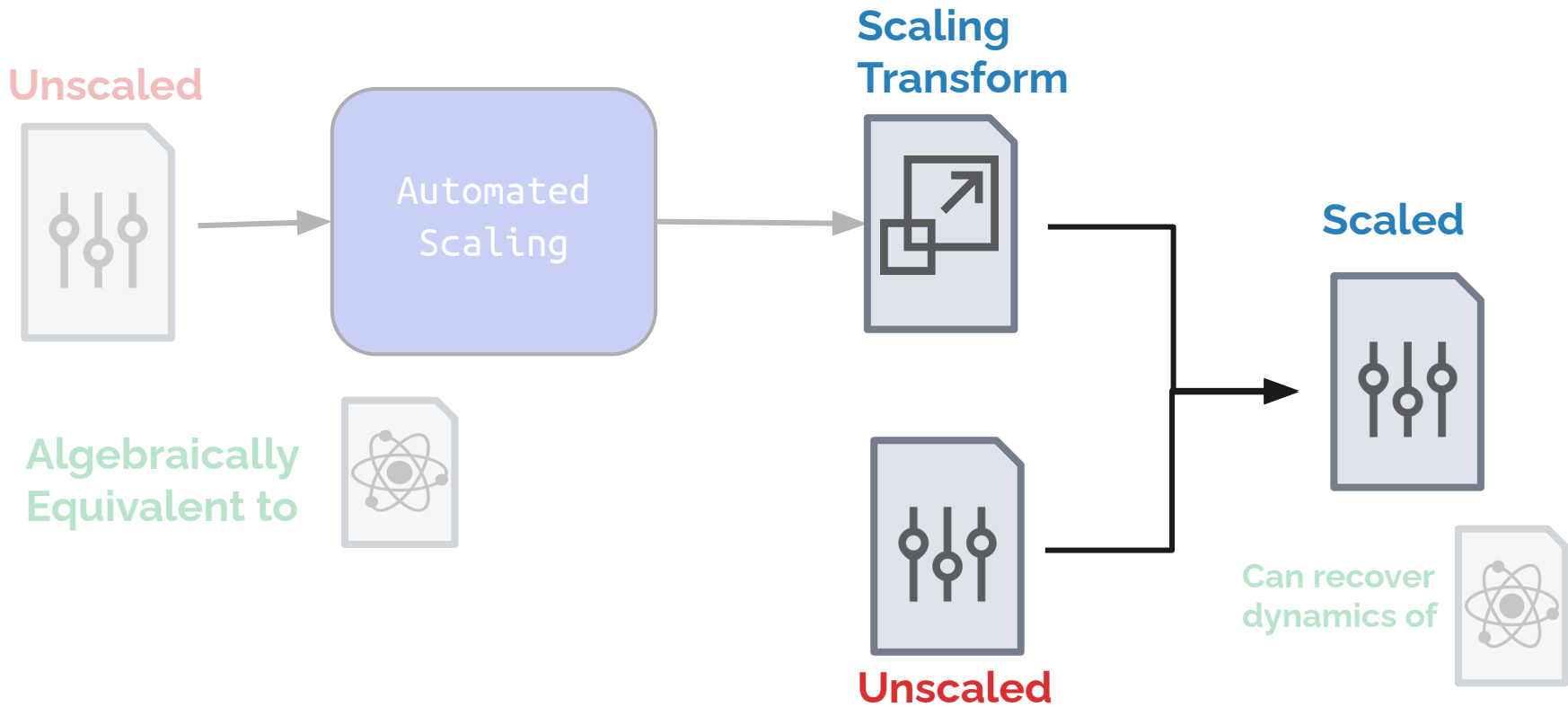
Unscaled



Can recover dynamics of



Automated Scaling



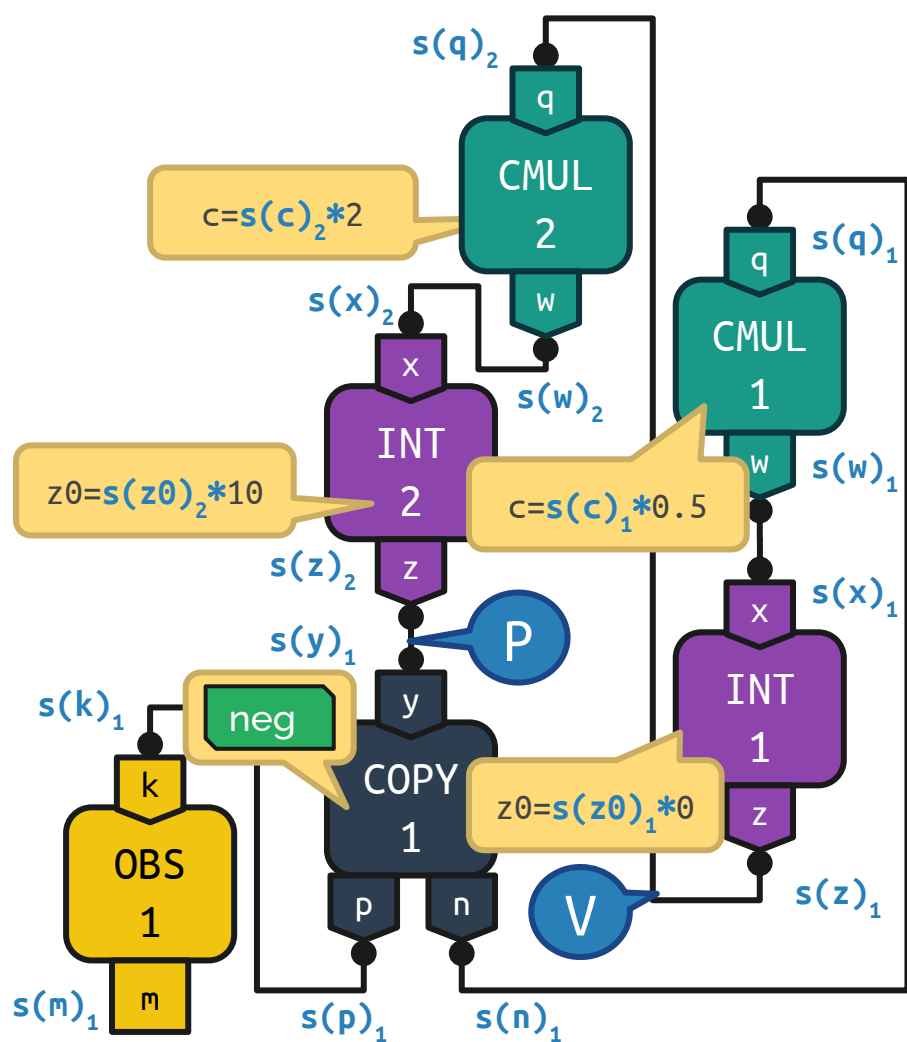
Automated Scaling

The Scaling Transform

$s(z)_2$ Signal scale factors

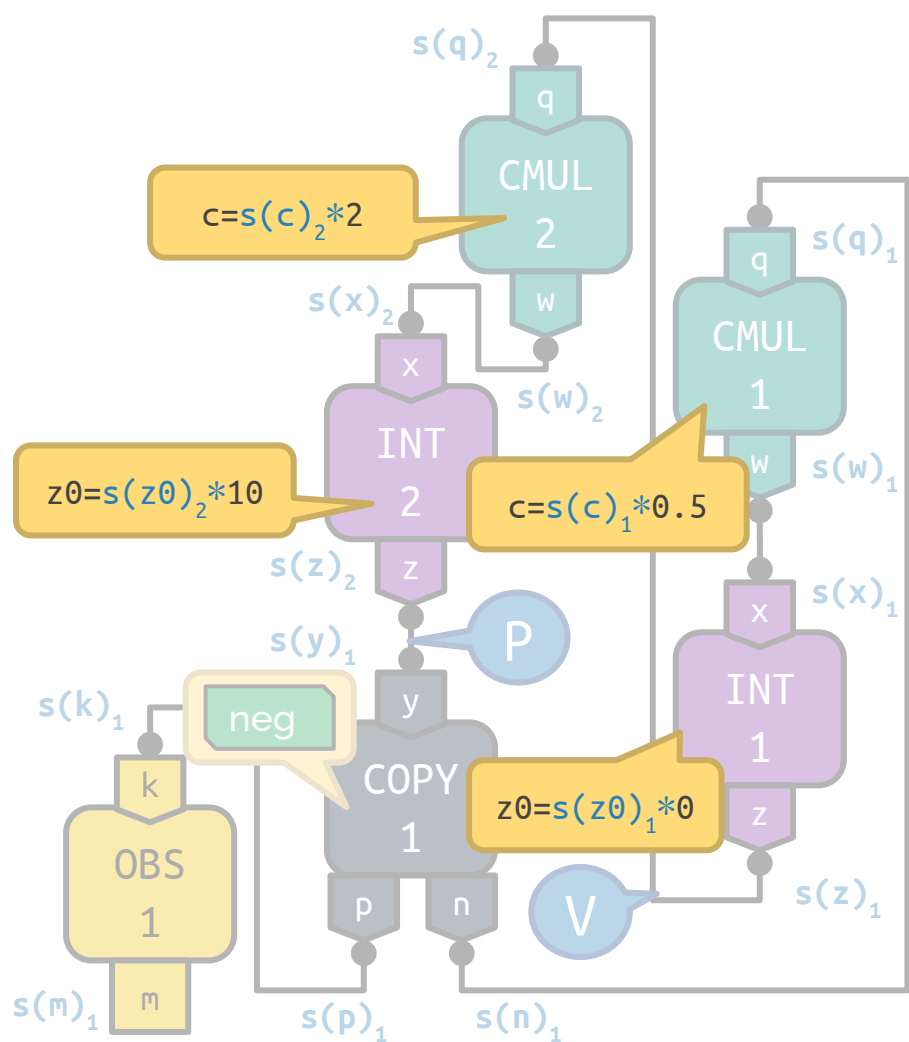
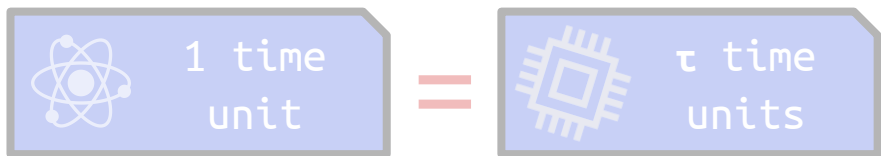
$s(c)_1$

τ Time scale factor



Applying the Scaling Transform

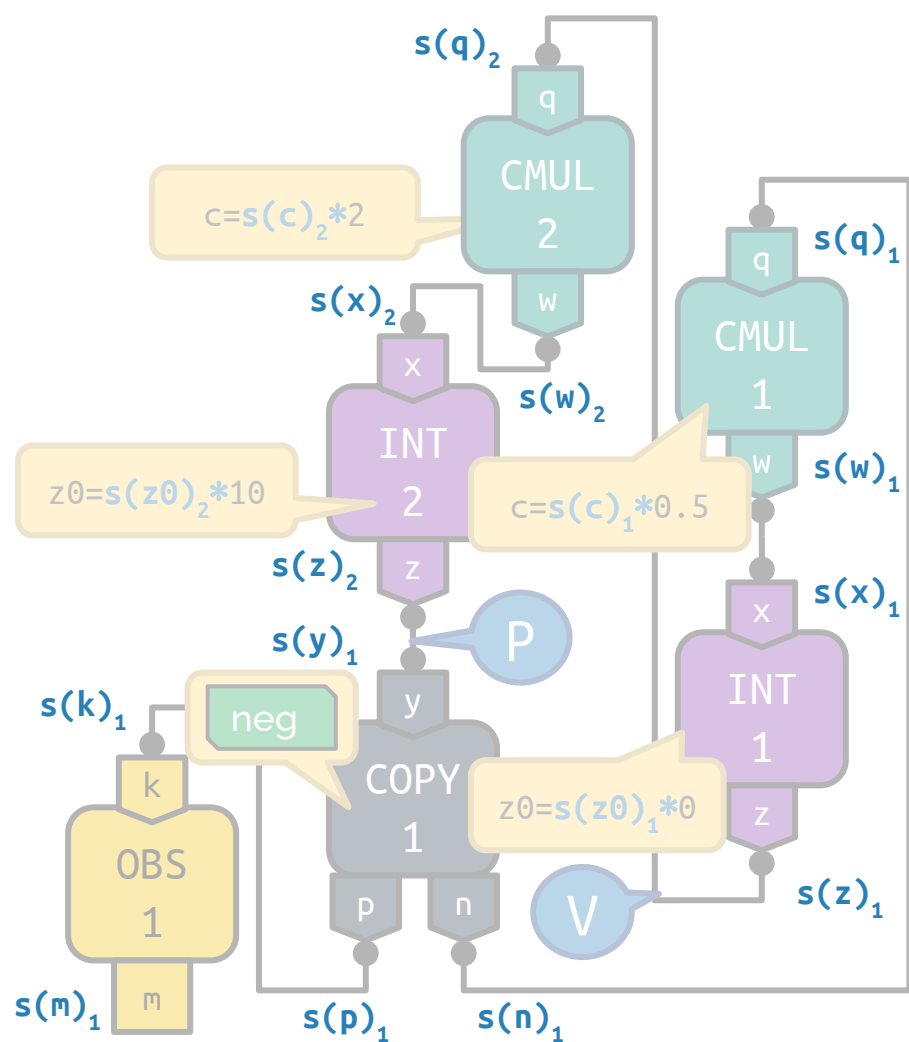
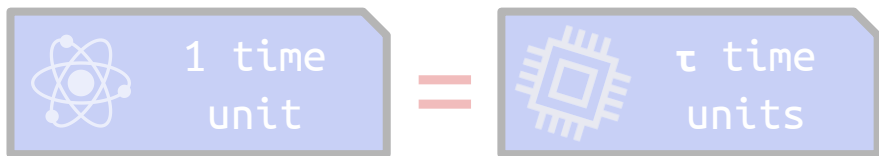
Multiply each digital value with its signal scale factor



Applying the Scaling Transform

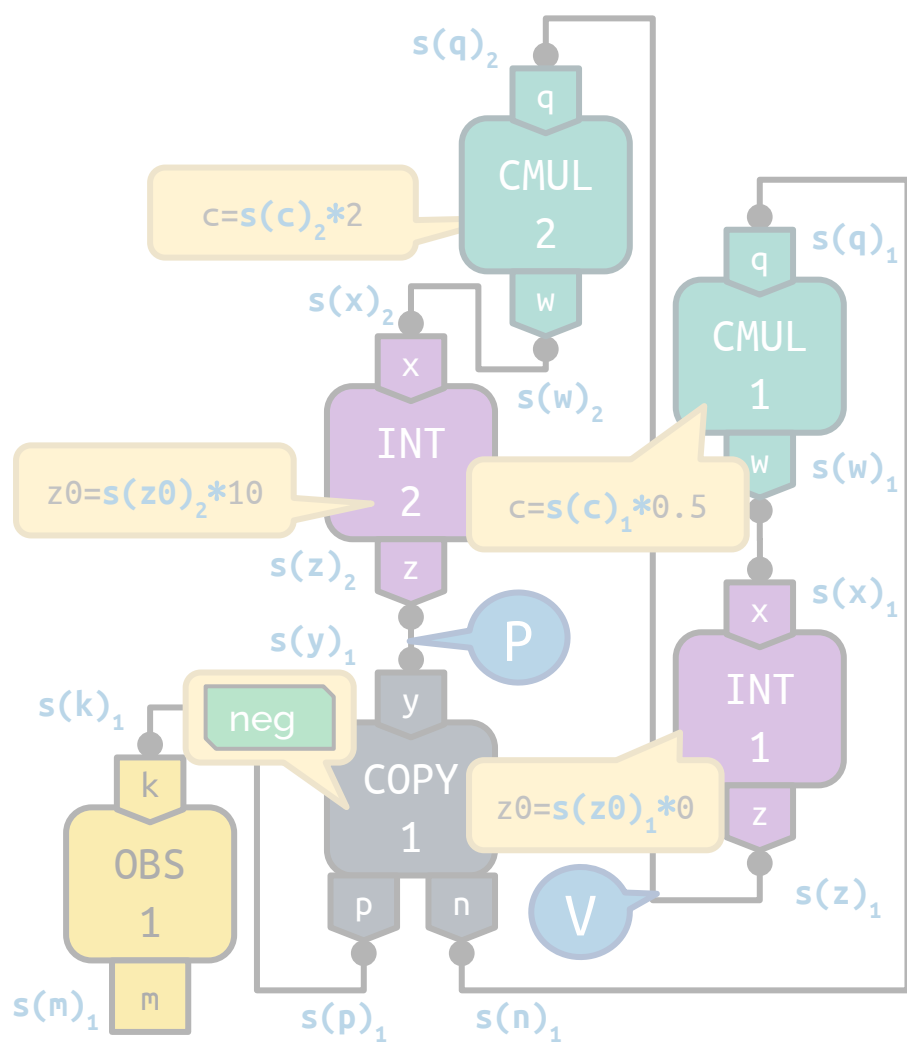
The scaled signals propagate through the analog device

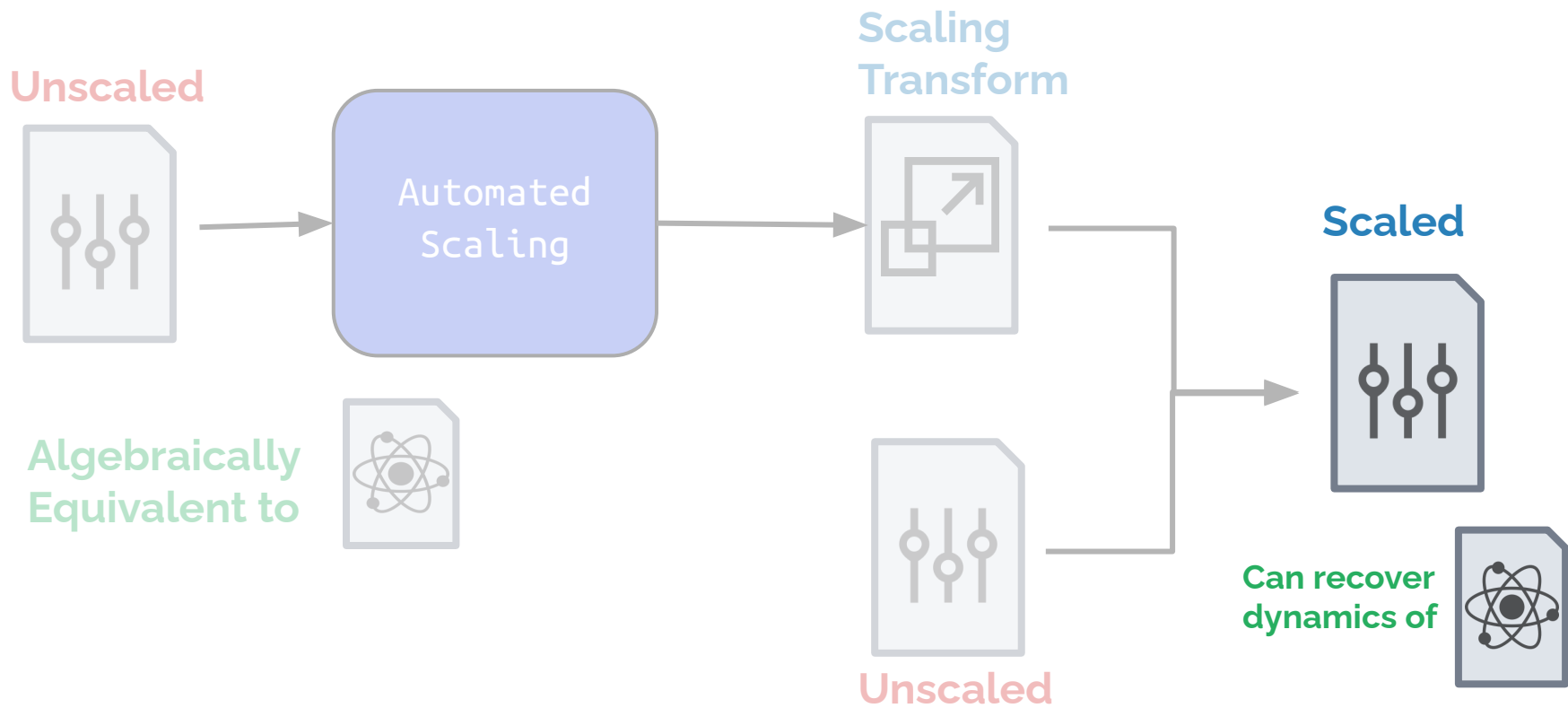
Respects physical restrictions



Applying the Scaling Transform

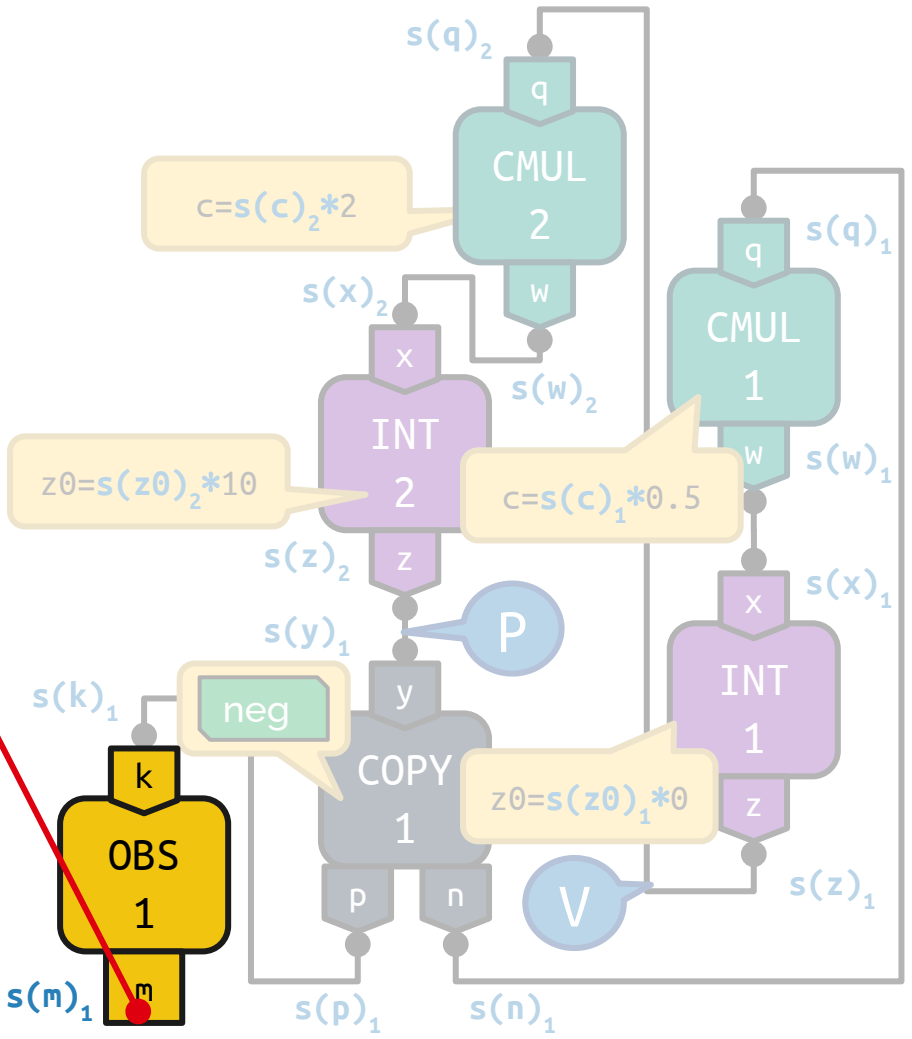
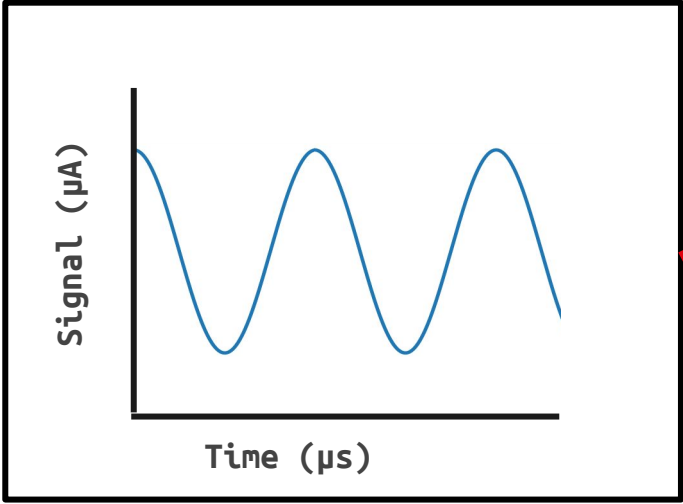
This automatically changes the execution speed of the dynamical system



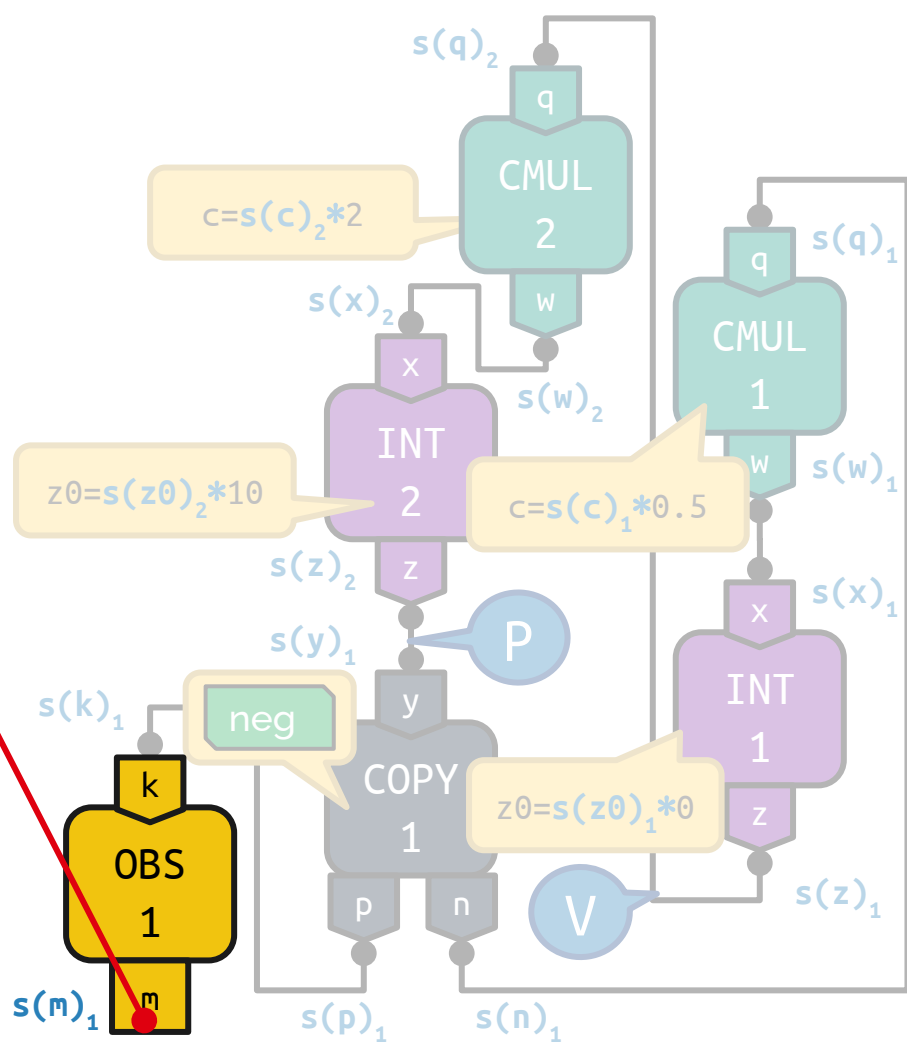
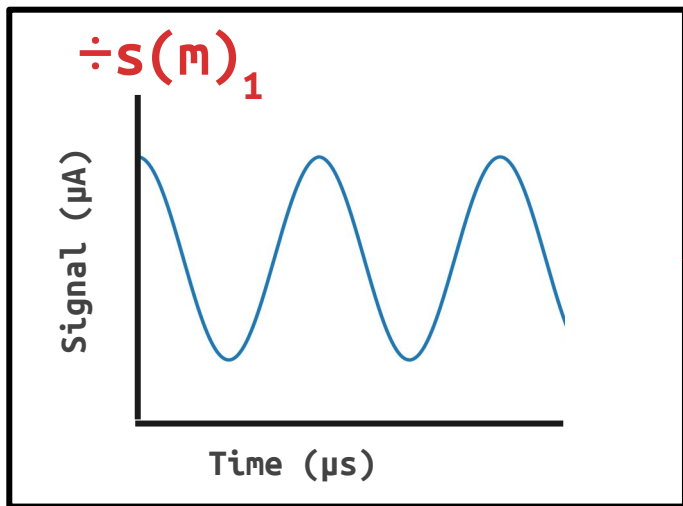


Automated Scaling

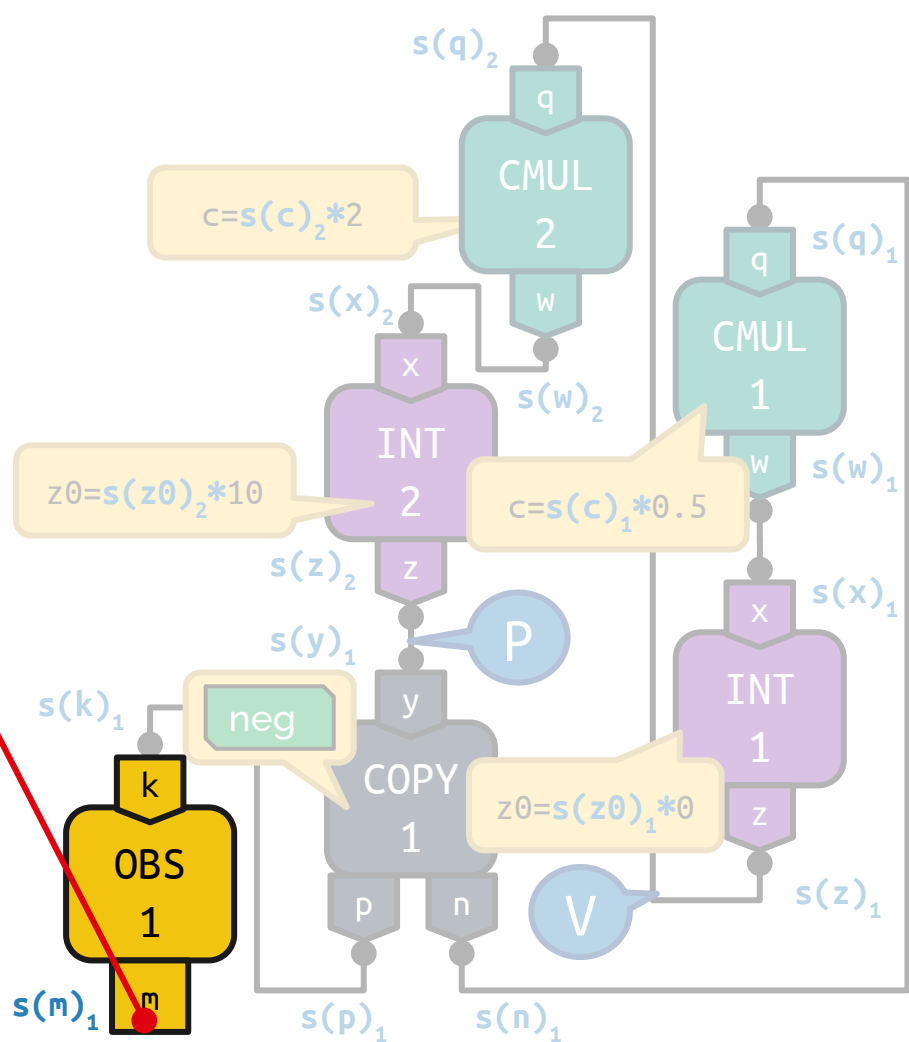
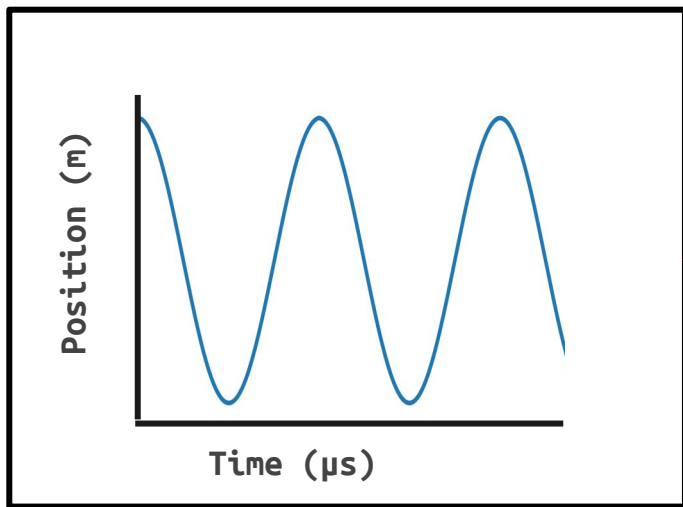
Recovering the Original Dynamics



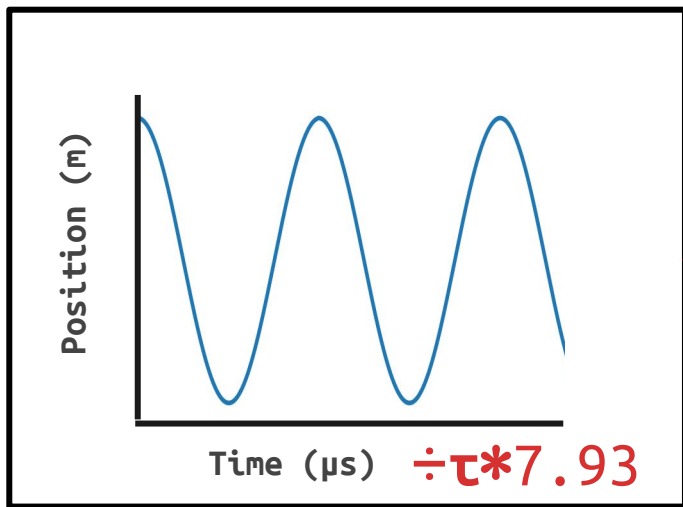
Recovering the Original Dynamics



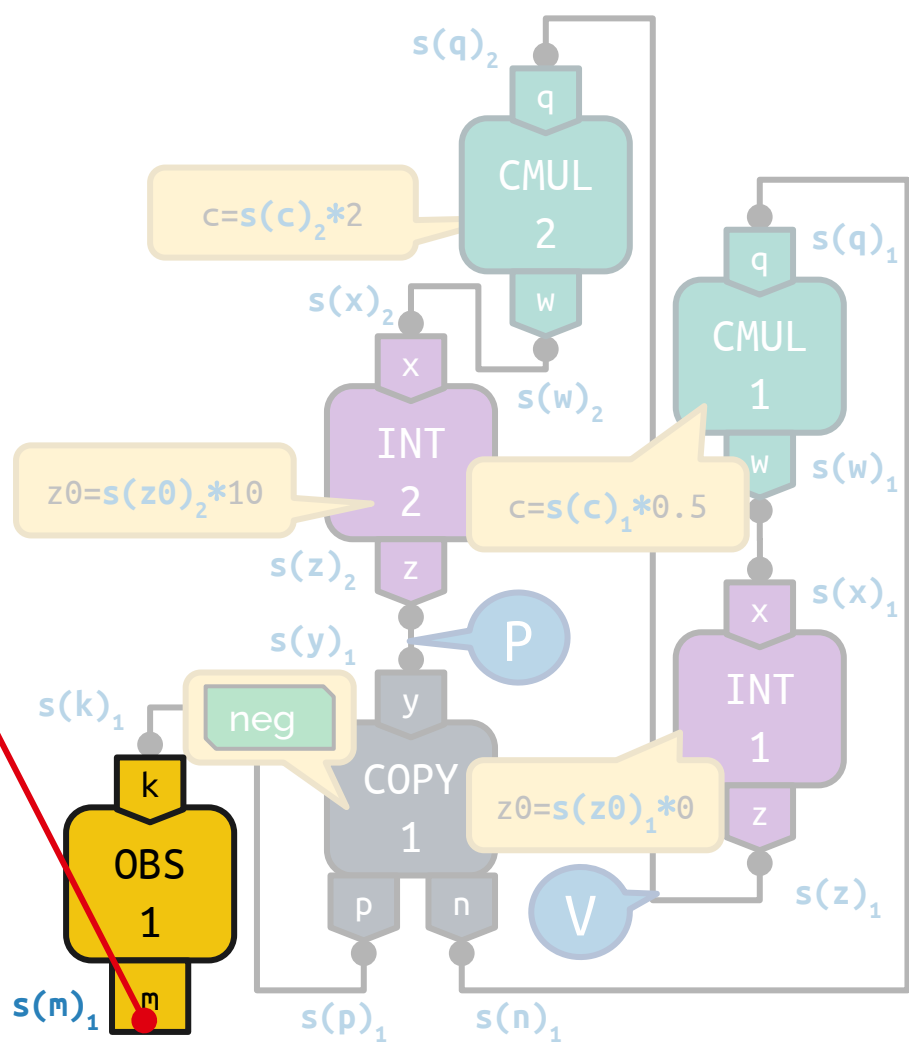
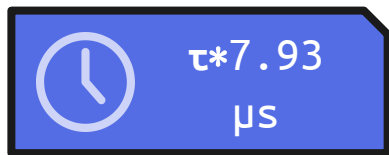
Recovering the Original Dynamics



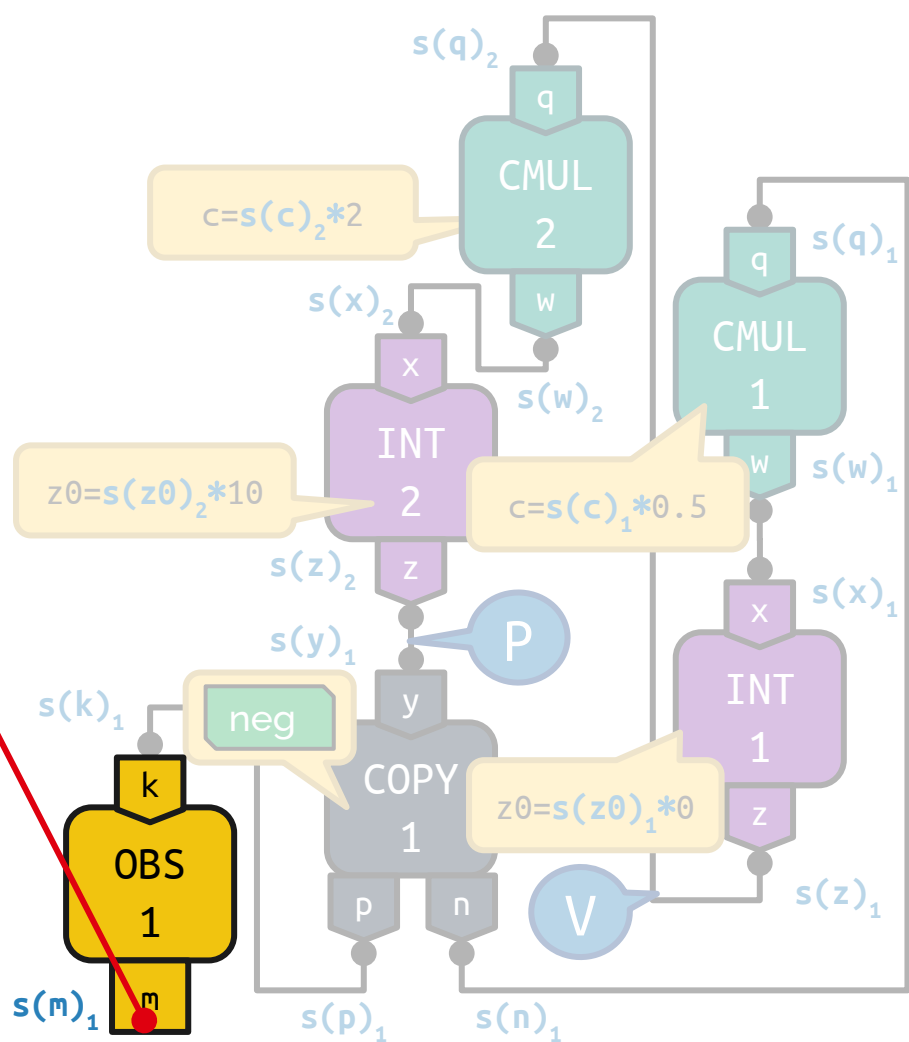
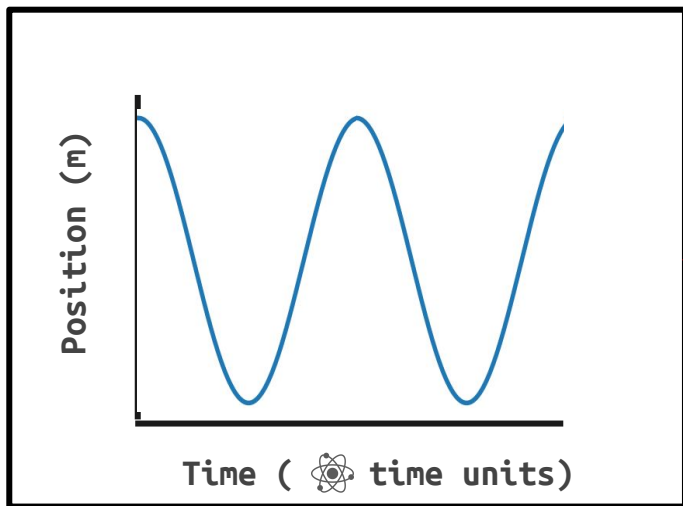
Recovering the Original Dynamics

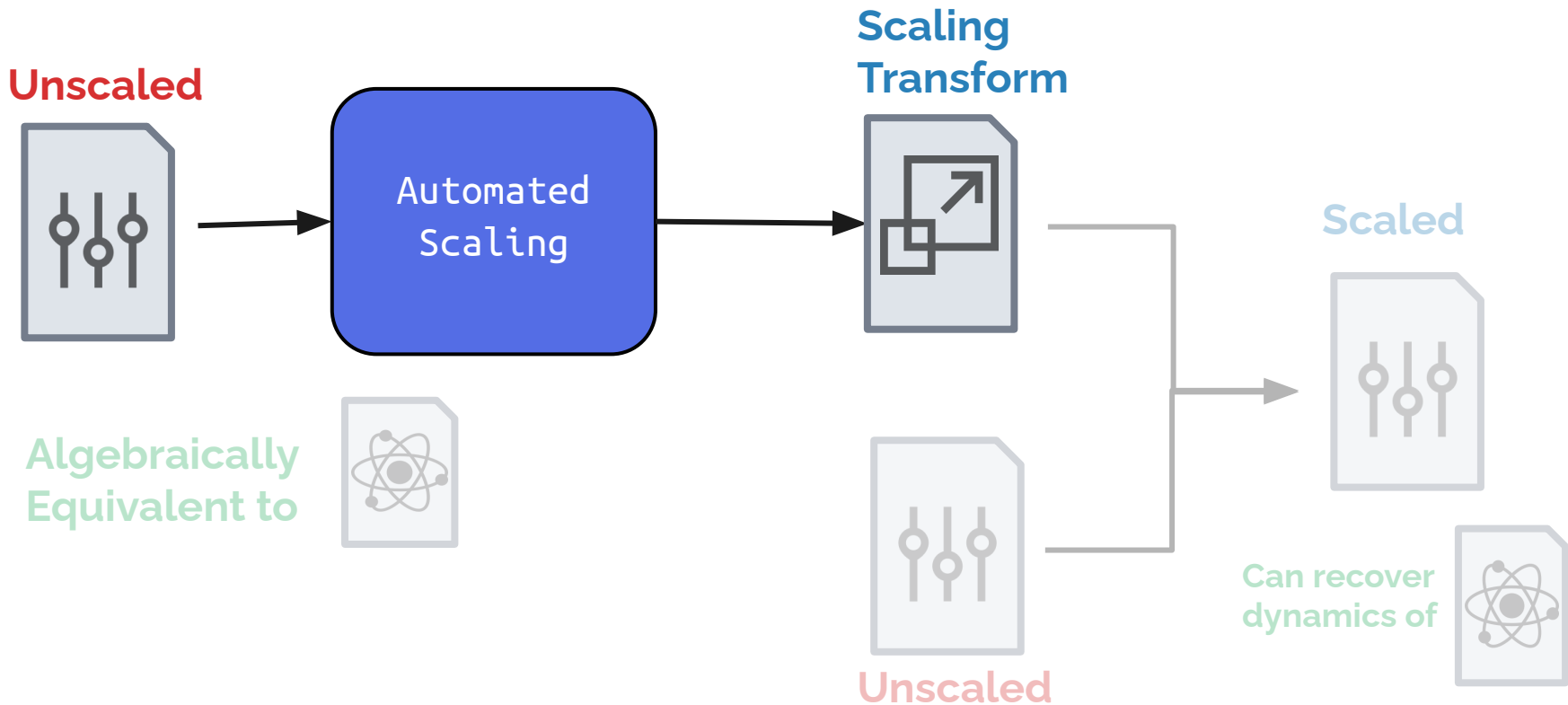


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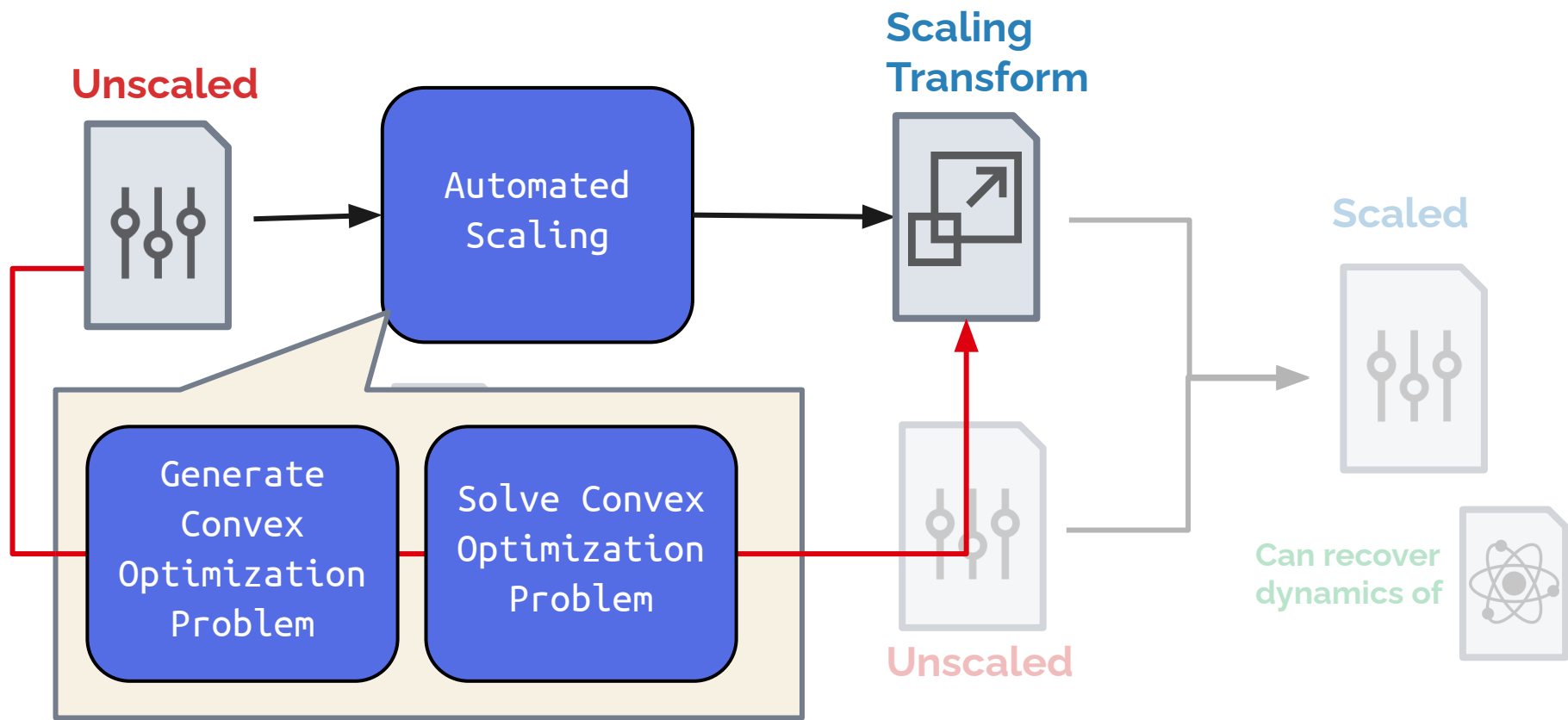


Recovering the Original Dynamics

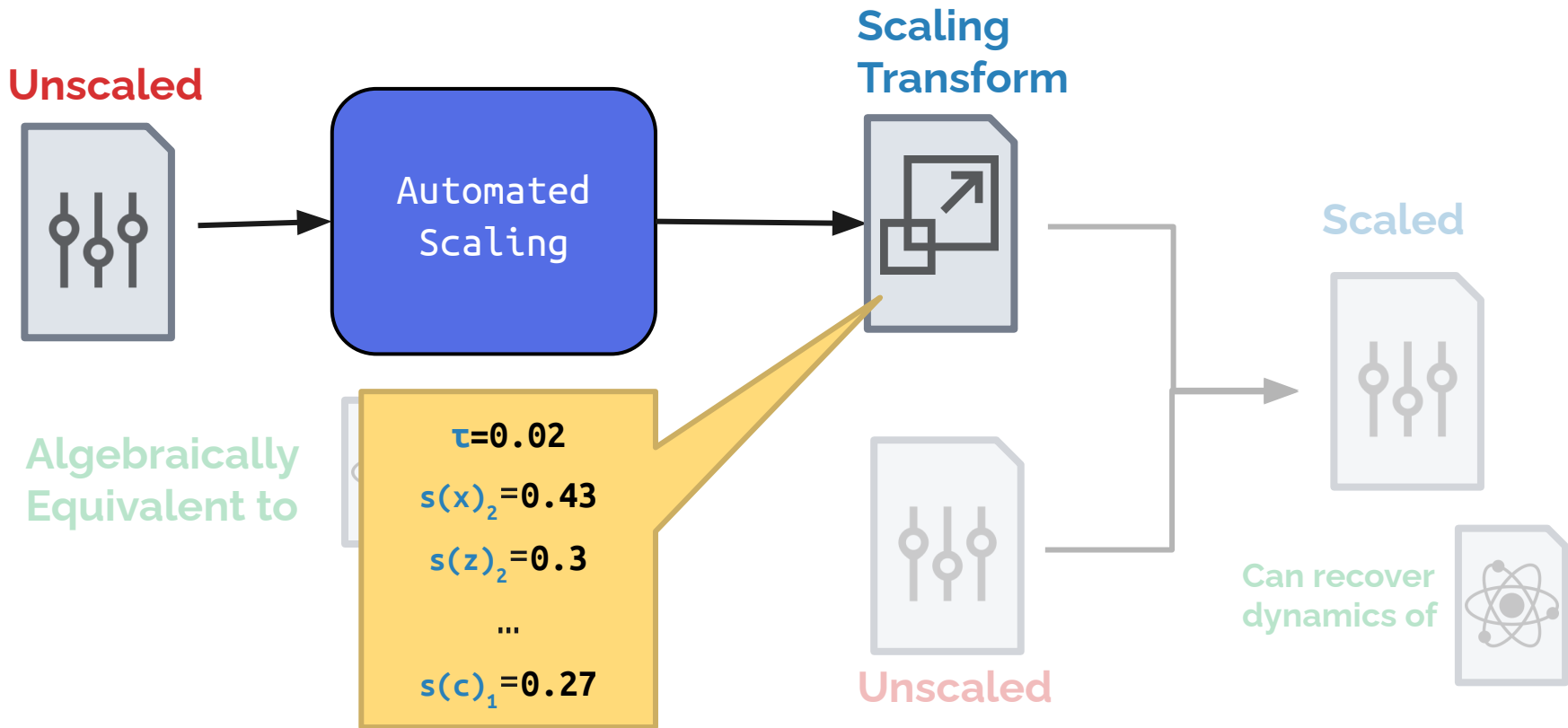




Automated Scaling



Automated Scaling



Automated Scaling

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

Automatic Scaling

Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

Operating Range Restrictions

Execution Speed Constraints

Execution Speed Restrictions

Quality Constraints

Noise and Quantization Error

Connection Constraints

Factor Constraints

Manufacturing Variations

Constraints encode physical limitations of analog device

Automatic Scaling

Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints



Constraints ensure original dynamical system is recoverable

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

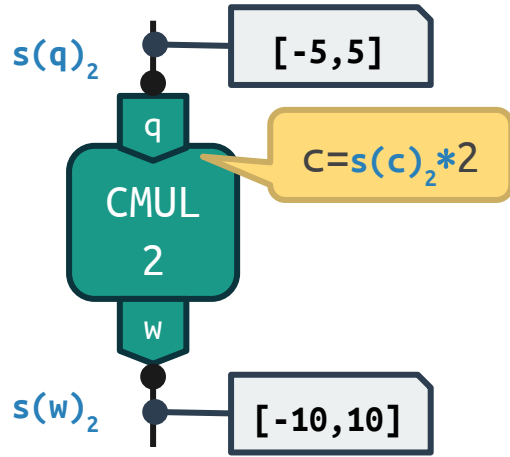
Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

Operating Range Constraints



$$q, w \in [-2, 2] \mu\text{A}$$
$$c \in [-1, 1]$$

$$s(q)_2 * [-5, 5] \subseteq [-2, 2] \mu\text{A}$$
$$s(w)_2 * [-10, 10] \subseteq [-2, 2] \mu\text{A}$$
$$s(c)_2 * 2 \subseteq [-1, 1]$$

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

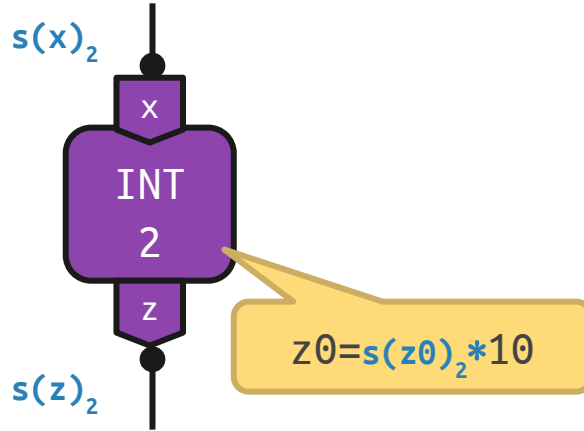
Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

Factor Constraints

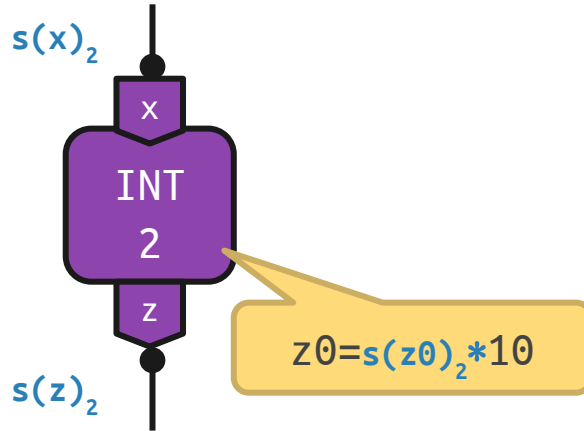


$$s(z)_2 * z = \int 0.93 * 0.5 * (s(x)_2 * x) \tau^{-1} dt_{hw}$$

$$s(z)_2 * z(0) = 0.77 * s(z0)_2 * 10$$

Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)

Factor Constraints

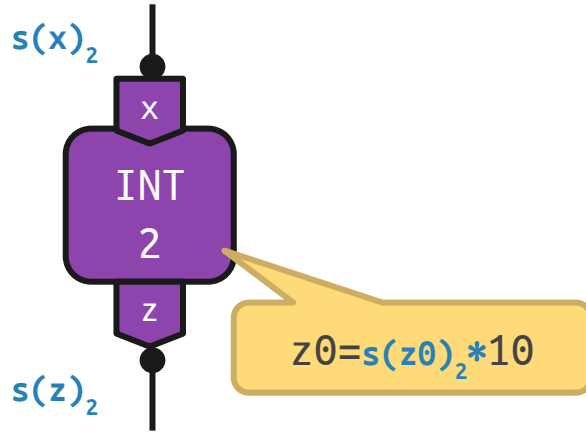


$$s(z)_2 * z = (0.93 * s(x)_2 * \tau^{-1}) \int 0.5 * (x) dt_{hw}$$

$$s(z)_2 * z(0) = (0.77 * s(z0)_2) 10$$

Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)

Factor Constraints



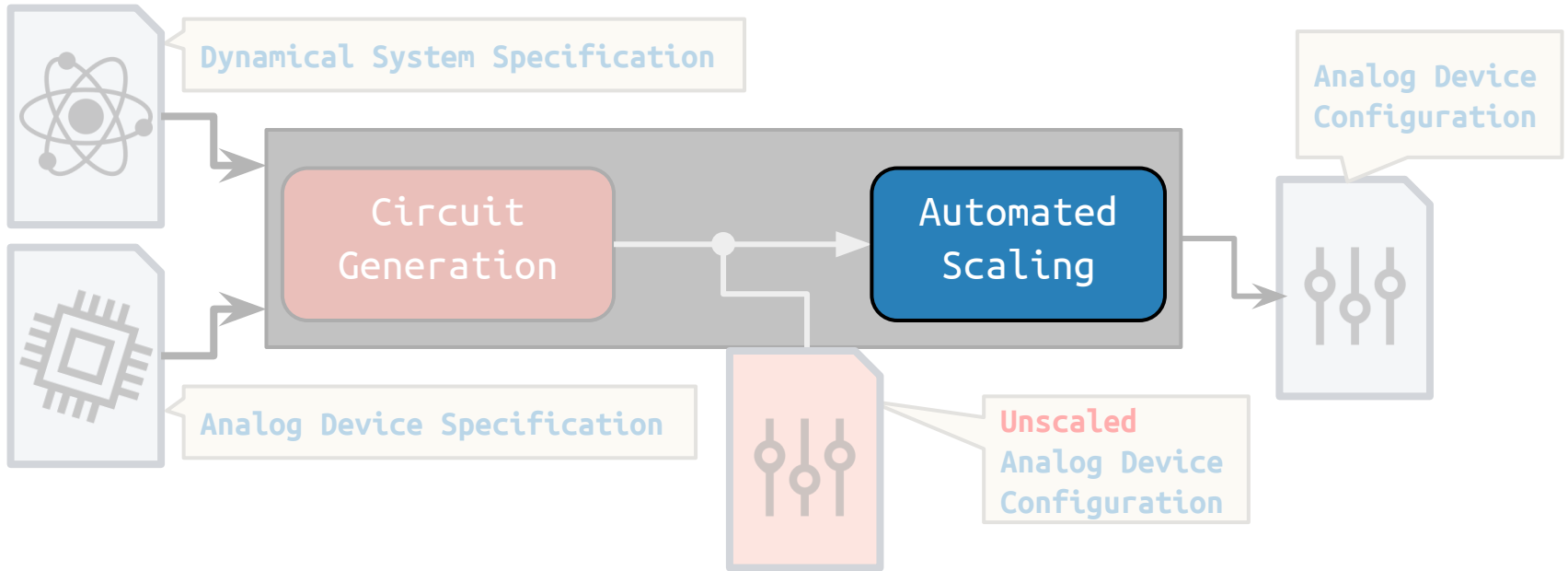
$$z = \int 0.5 * (x) dt_{hw}$$

$$s(z)_2 = 0.93 * s(x)_2 * \tau^{-1}$$

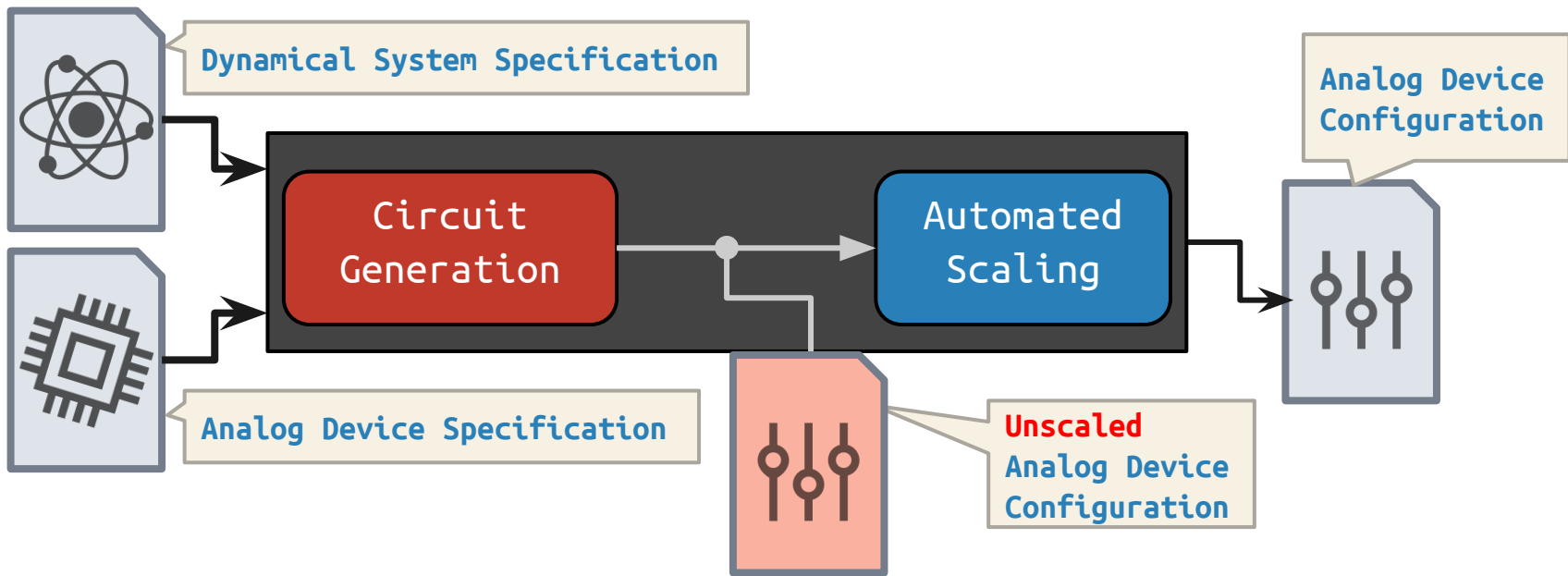
$$z(0) = 10$$

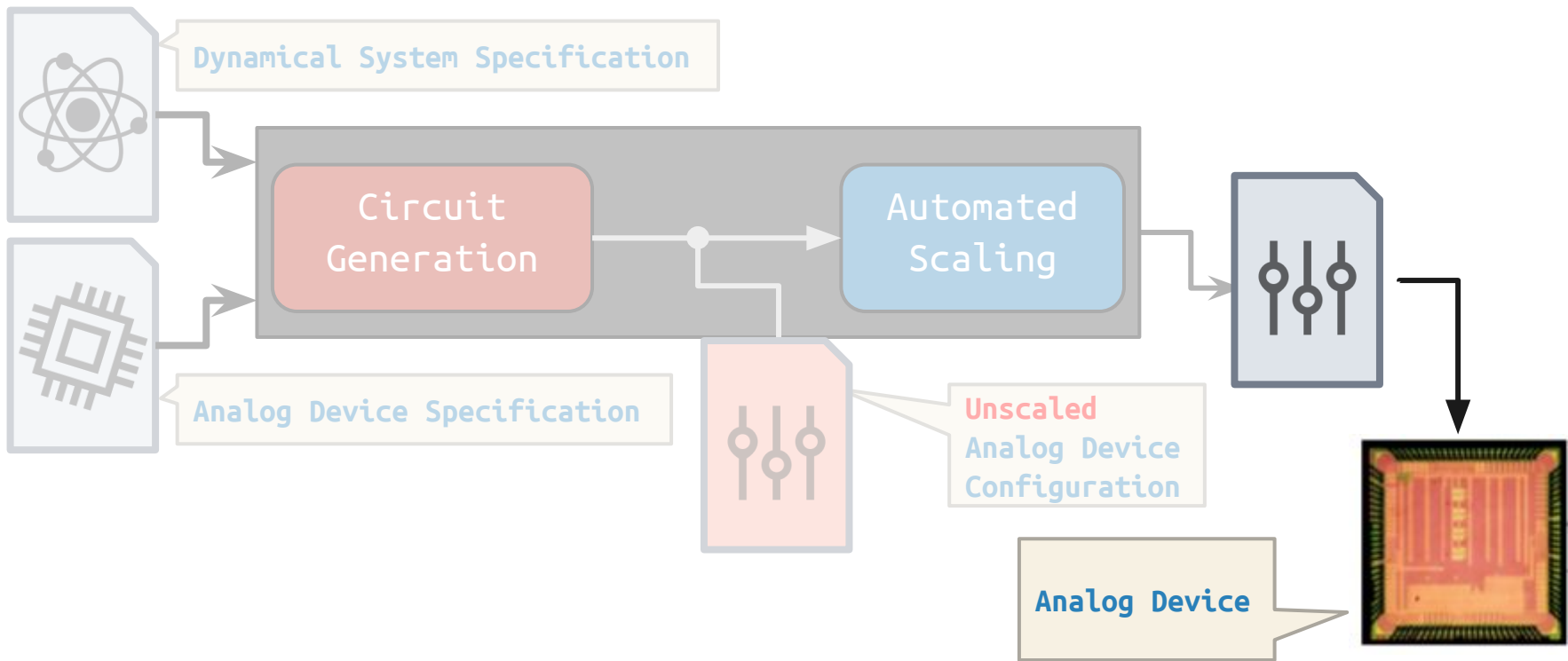
$$s(z)_2 = 0.77 * s(z0)_2$$

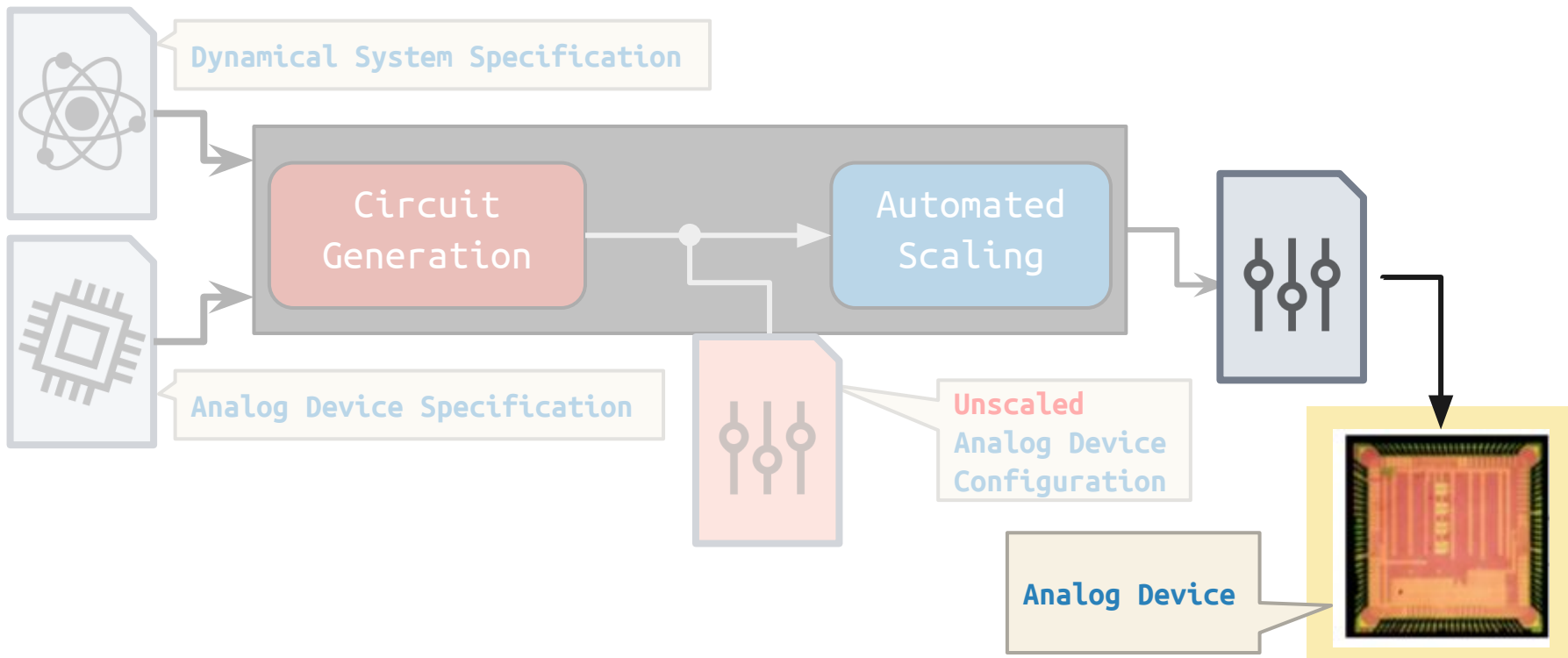
Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)



1. Time Dilation and Contraction for Programmable Analog Devices with Jaunt. ASPLOS 2018.
2. Noise-Aware Dynamical System Compilation for Analog Devices with Legno. ASPLOS 2020.



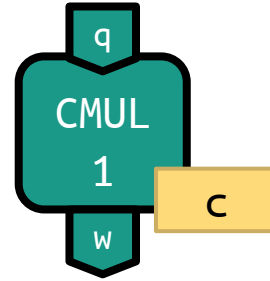
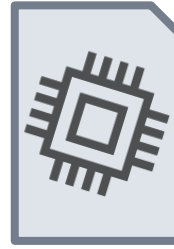




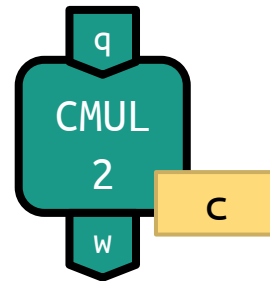
Inputs



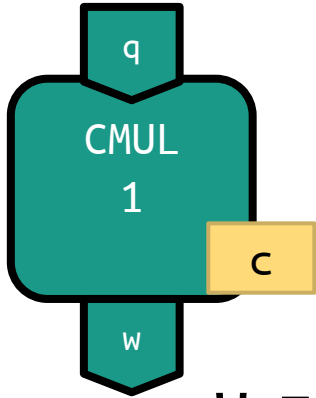
Manufacturing Variations



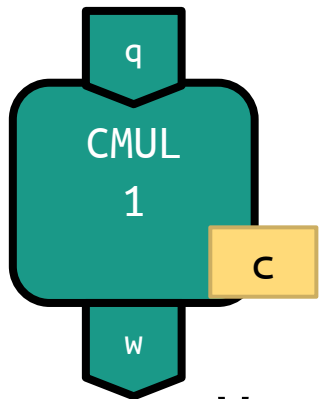
$$w = 0.86 * c * q$$



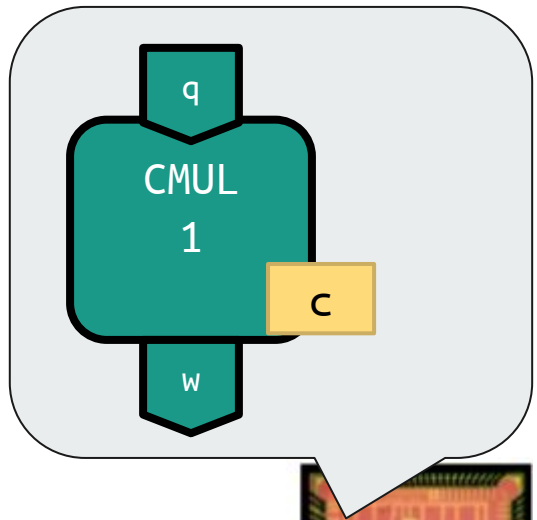
$$w = 1.11 * c * q$$



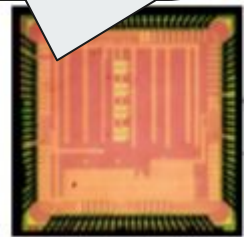
$$w = c * q$$

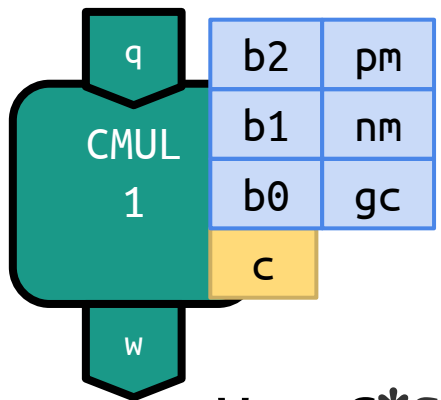


$$w = c * q$$

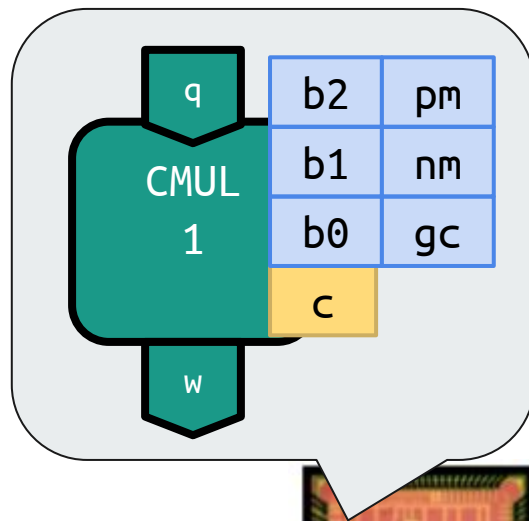


$$w = f(c, q)$$

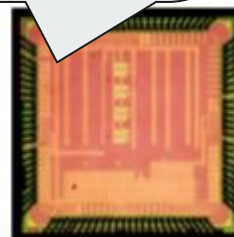


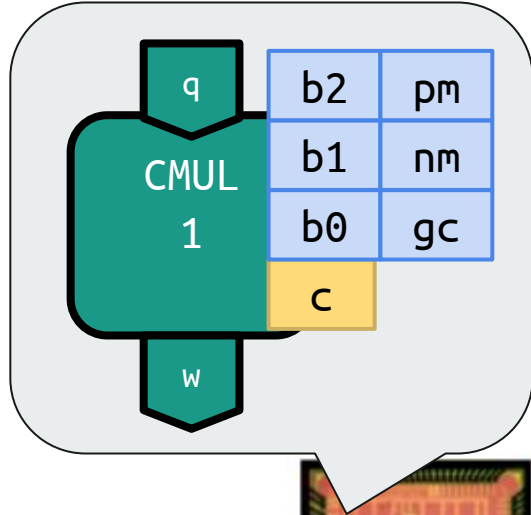


$$w = c * q$$

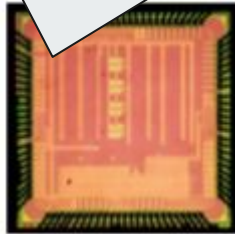


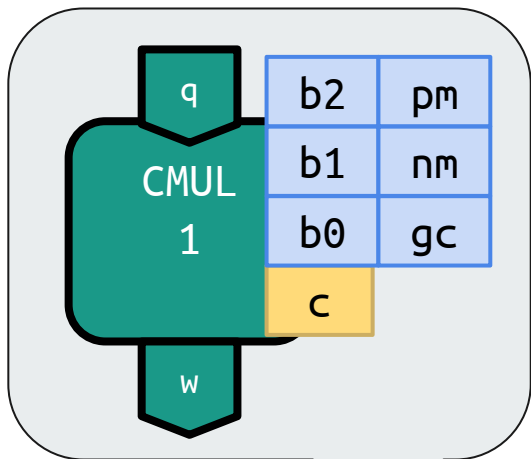
$$w = f(c, q)$$



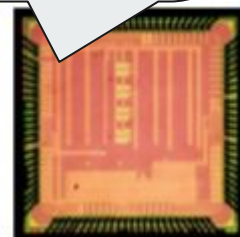
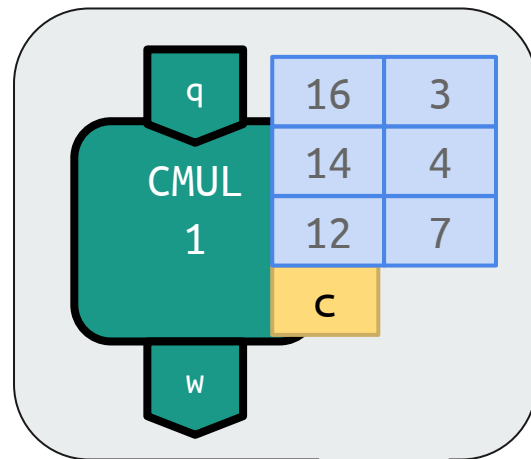
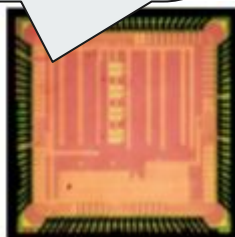


$$w = f(c, q)$$

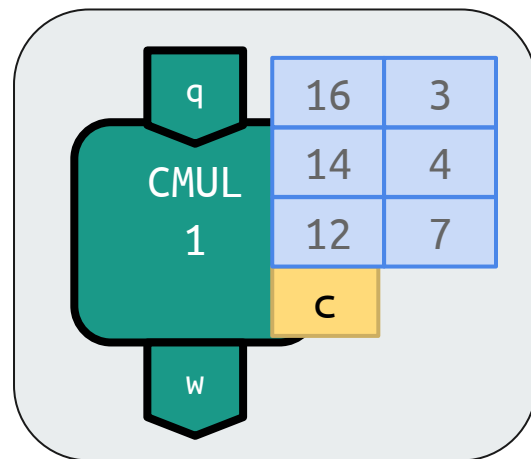
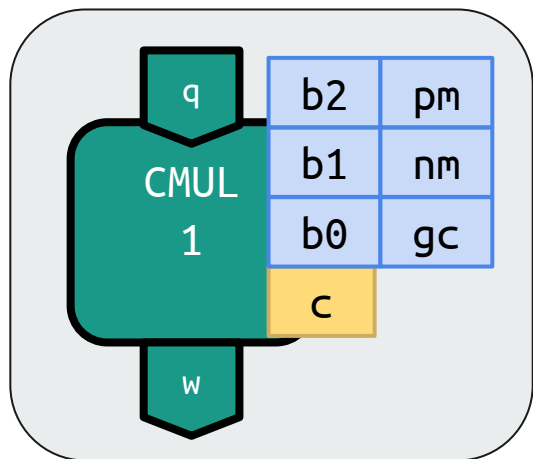




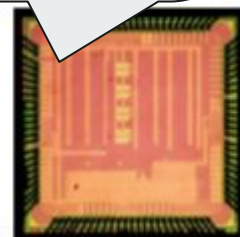
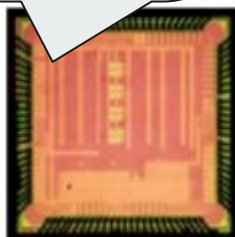
$$w = f(c, q)$$



Objective: Minimize the error between expected, observed dynamics

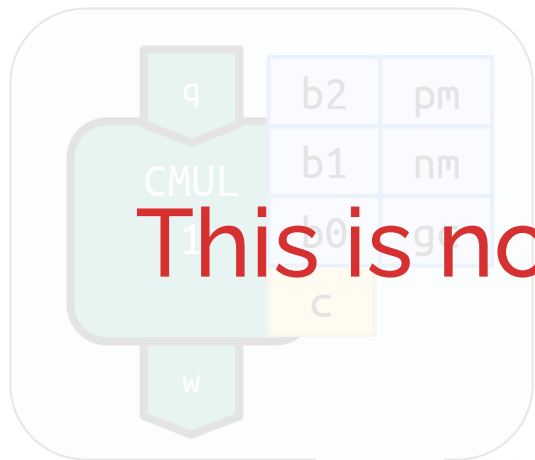


$$w = f(c, q)$$

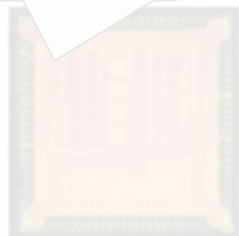


$$\text{minimize } \|f(c, q) - c * q\|$$

Objective: Minimize the error between expected, observed dynamics

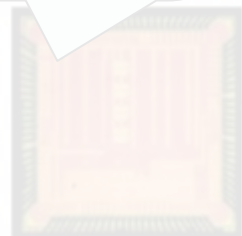


$$w = f(c, q)$$

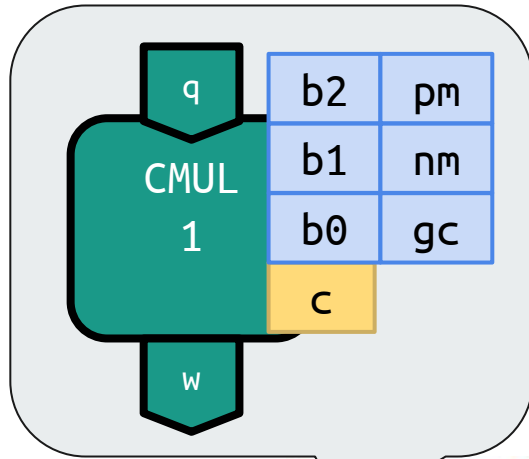


minimize $\|f(c, q) - c * q\|$

This is not the best calibration objective



Objective: Minimize the error between expected, observed dynamics

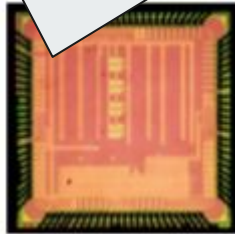


argmin

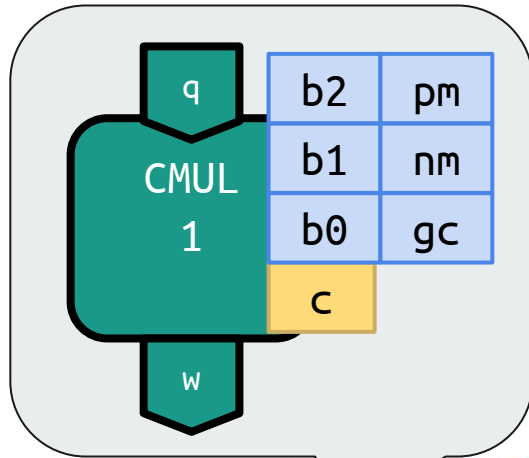
b2, b1, b0,
pm, nm, gc

$$\|f(c, q) - c * q\|$$

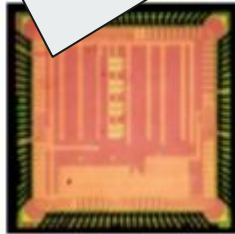
$$w = f(c, q)$$



Objective: Minimize the error between expected, observed dynamics



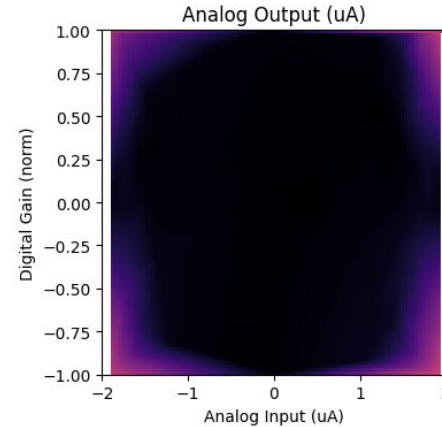
$$w = f(c, q)$$



argmin

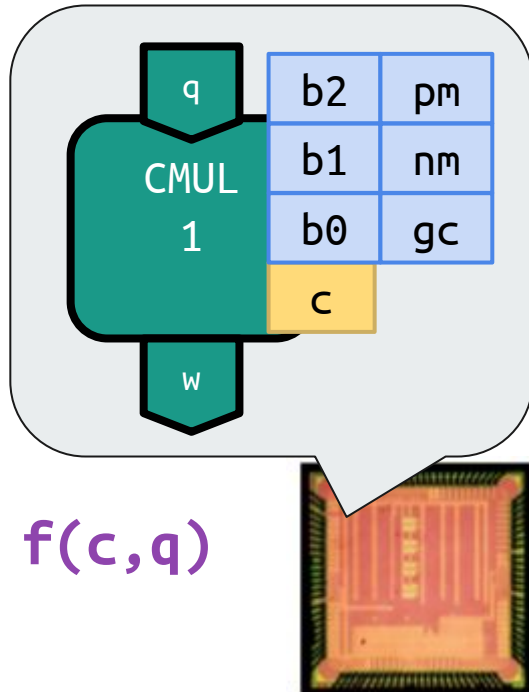
$b_2, b_1, b_0,$
 pm, nm, gc

$$\|f(c, q) - c * q\|$$



$$|f(c, q) - c * q|$$

Objective: Minimize the error between expected, observed dynamics

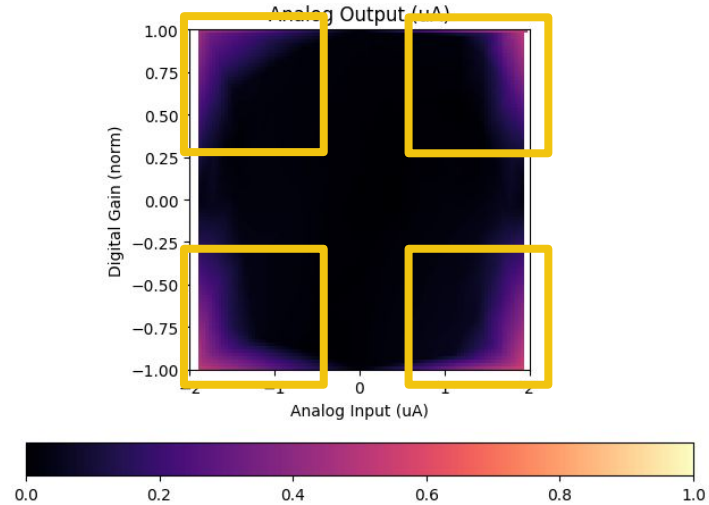


$$w = f(c, q)$$

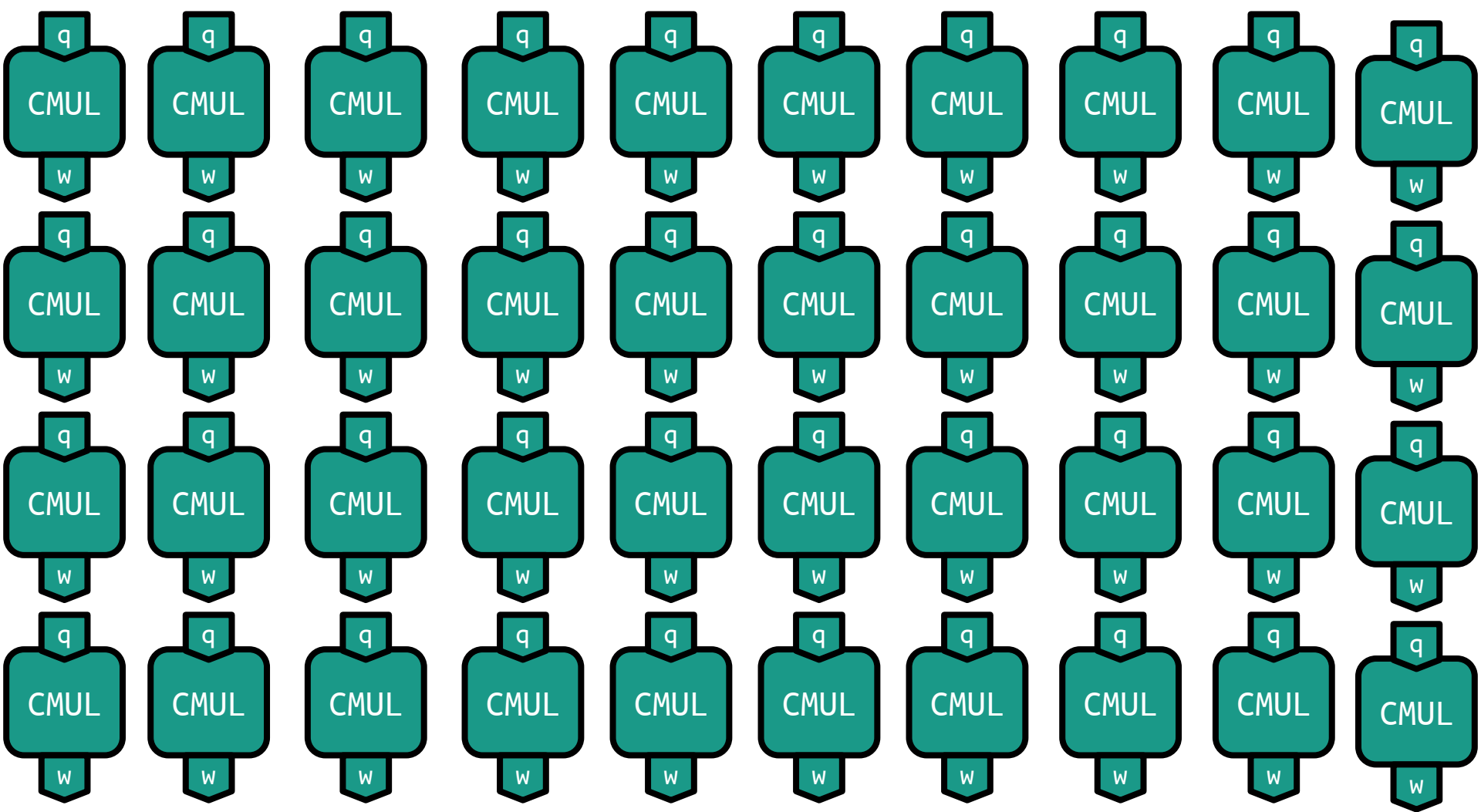
argmin

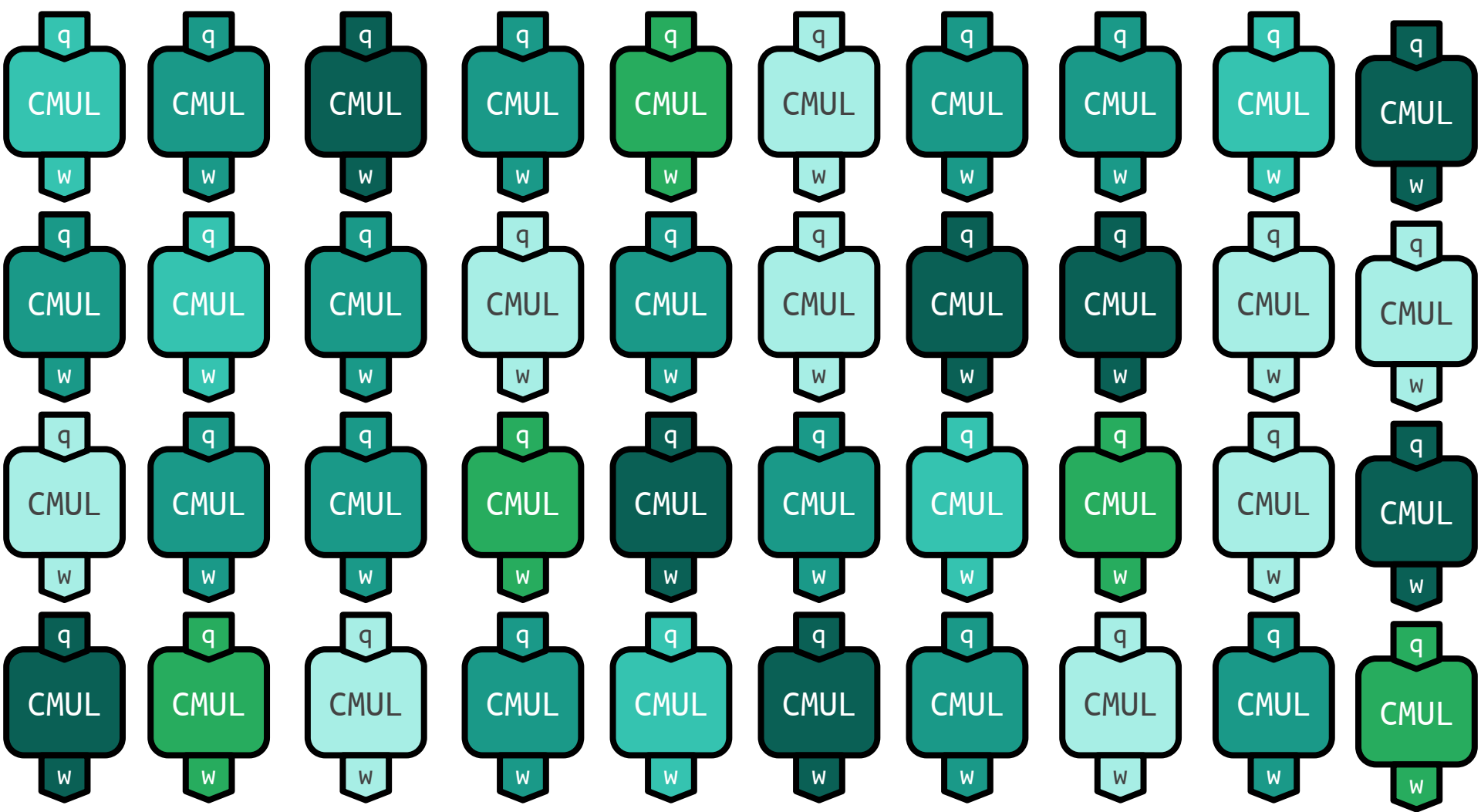
b2, b1, b0,
pm, nm, gc

$$\|f(c, q) - c * q\|$$



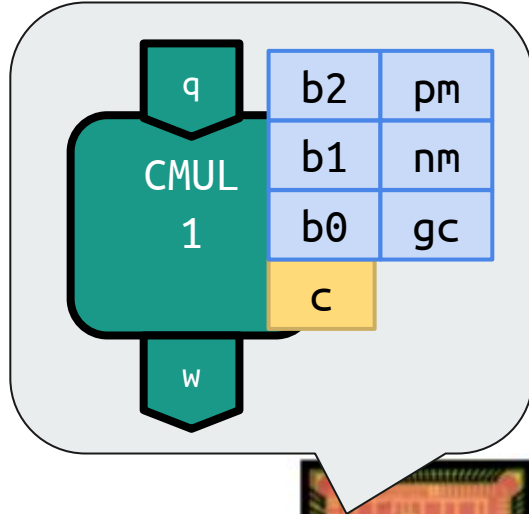
$$|f(c, q) - c * q|$$



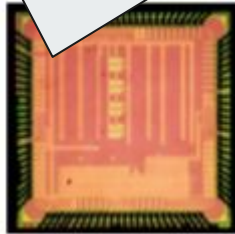


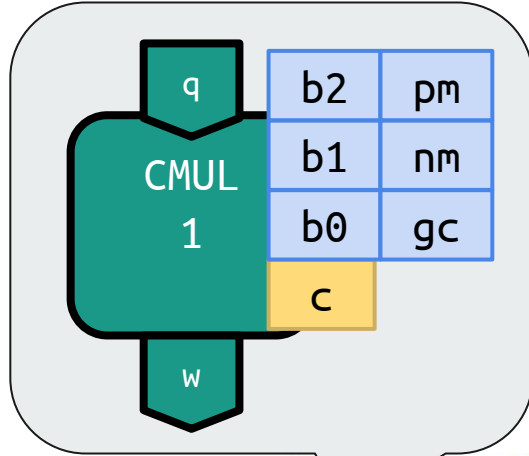
We allow calibrated multipliers to scale the result

$$w = \alpha * c * q$$

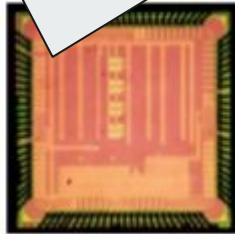


$$w = f(c, q)$$





$$w = f(c, q)$$



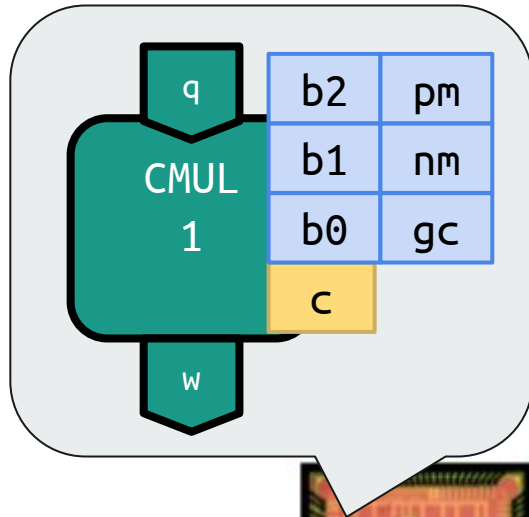
We allow calibrated multipliers to scale the result

$$w = \alpha * c * q$$

α is an empirically derived parameter

$$\operatorname{argmin}_{\alpha} \| f(c, q) - \alpha * c * q \|$$

Objective: Find the codes that produce the best fit for α

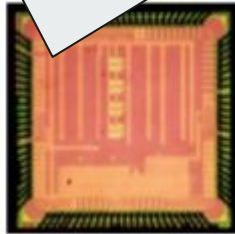


argmin

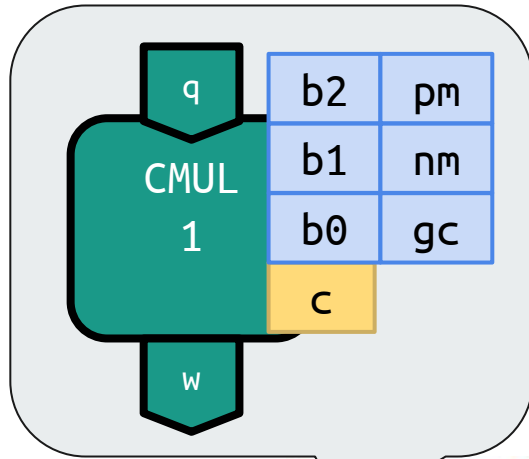
$b_2, b_1, b_0,$
 pm, nm, gc

$$\| f(c, q) - \alpha * c * q \|^2$$

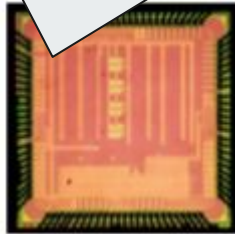
$$w = f(c, q)$$



Objective: Find the codes that produce the best fit for α



$$w = f(c, q)$$



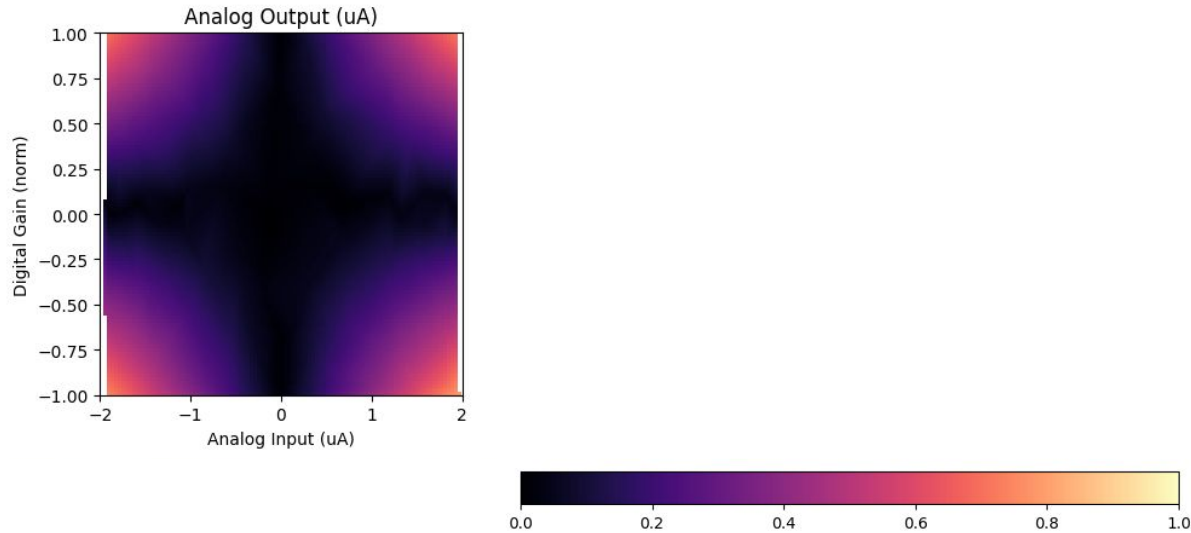
$$\operatorname{argmin}_{b_2, b_1, b_0, pm, nm, gc} \| f(c, q) - \alpha * c * q \|$$

For each combination of calibration codes, we find the best α

$$\operatorname{argmin}_{\alpha} \| f(c, q) - \alpha * c * q \|^2$$

Calibration Objective

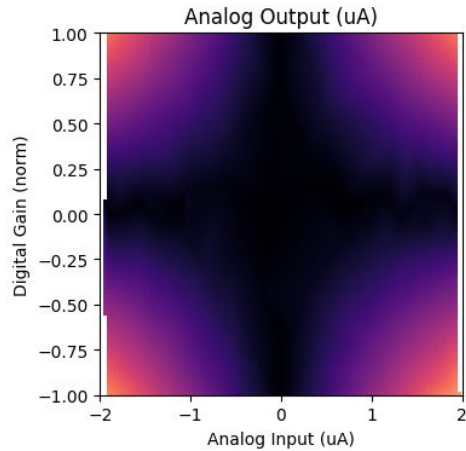
Best fitting α



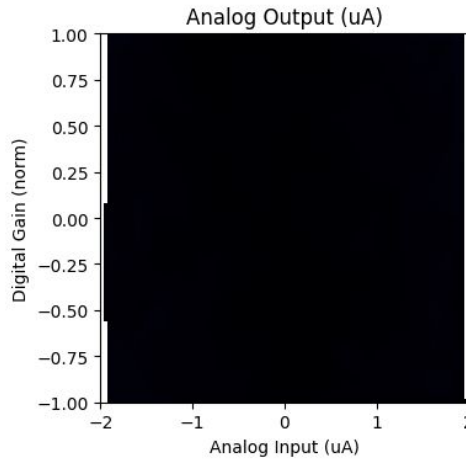
$$|f(c, q) - c * q|$$

Calibration Objective

Best fitting α



Best fitting α

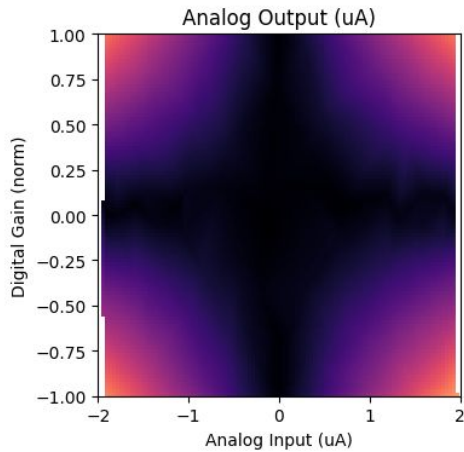


$$|f(c, q) - c * q|$$

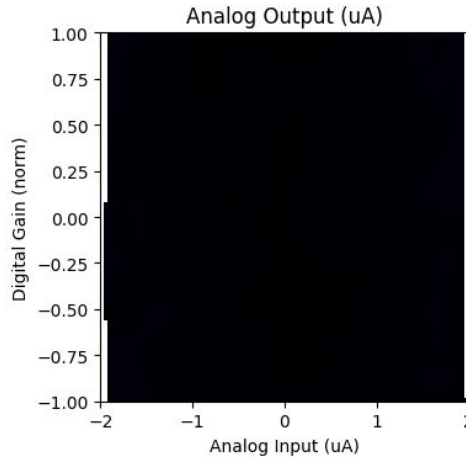
$$|f(c, q) - \alpha * c * q|$$

Calibration Objective

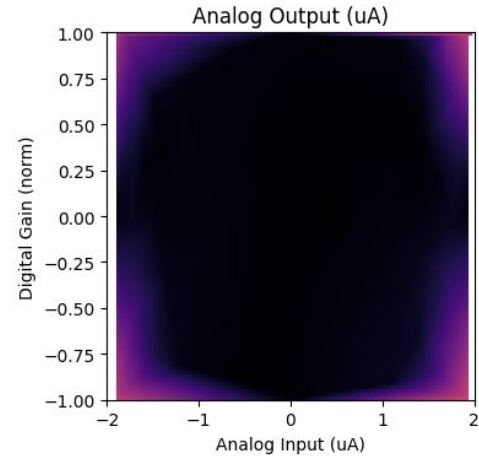
Best fitting α



Best fitting α



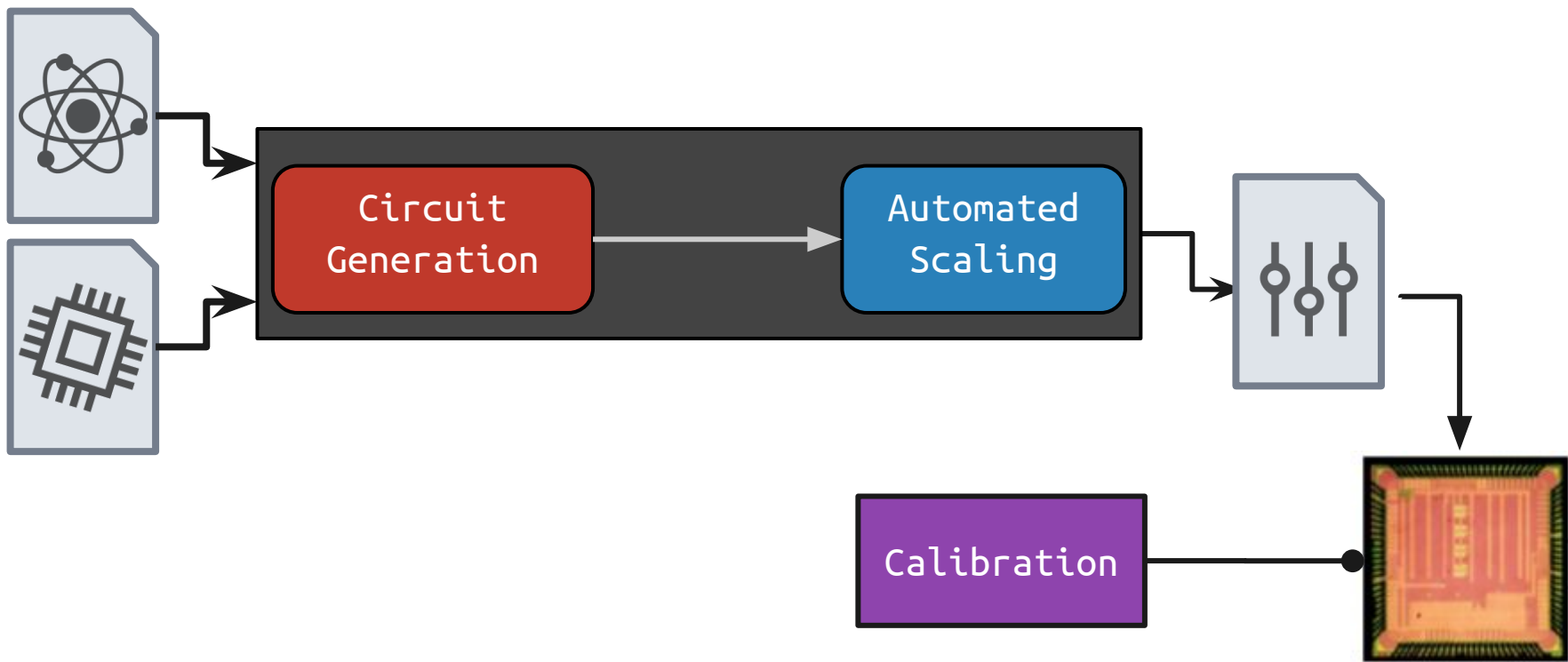
Minimize error



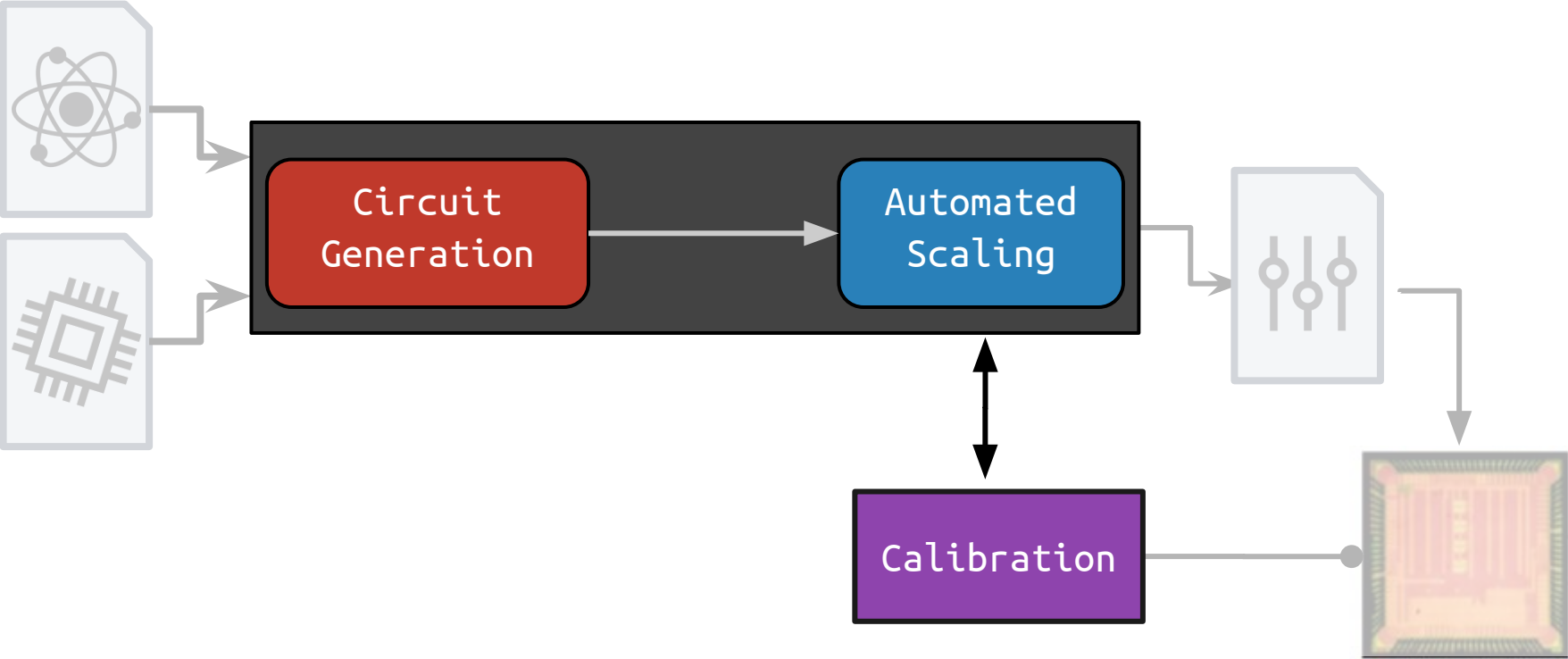
$$|f(c, q) - c * q|$$

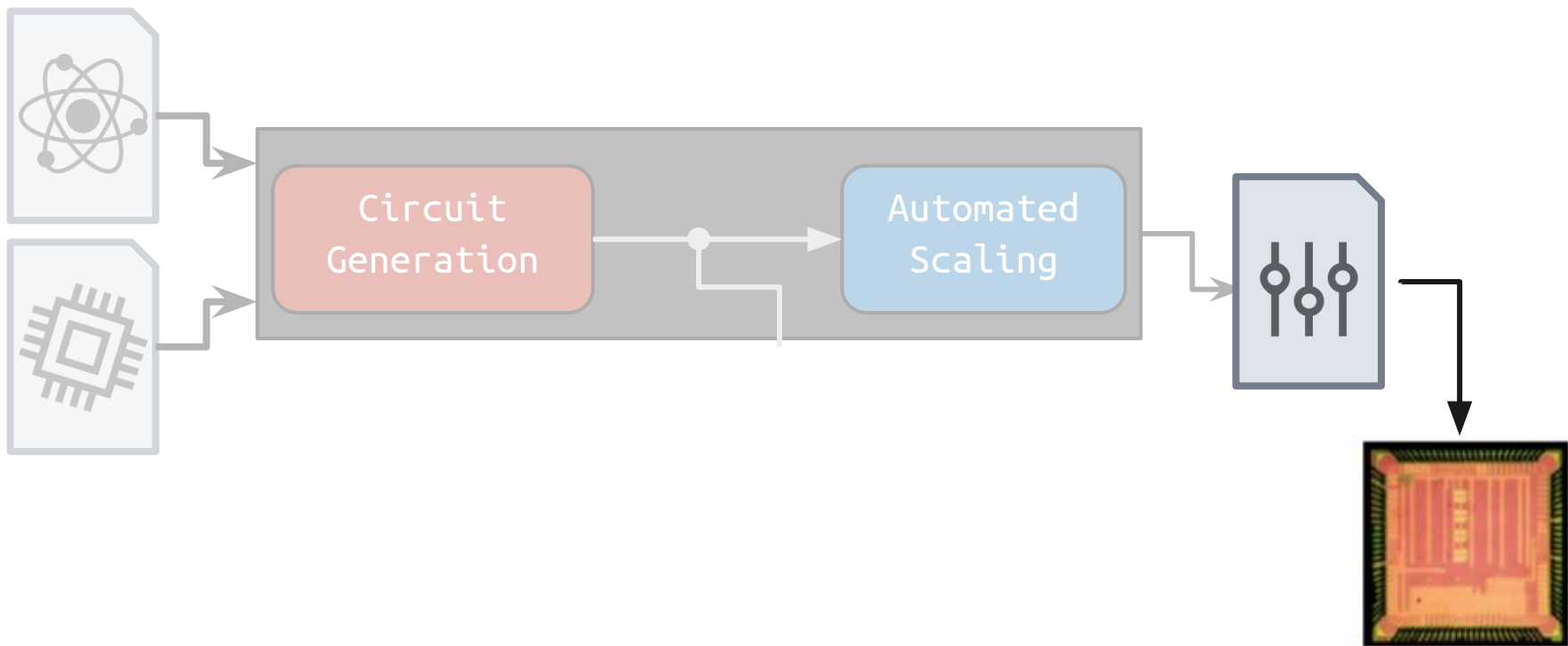
$$|f(c, q) - \alpha * c * q|$$

$$|f(c, q) - c * q|$$



Synergy between compilation, calibration





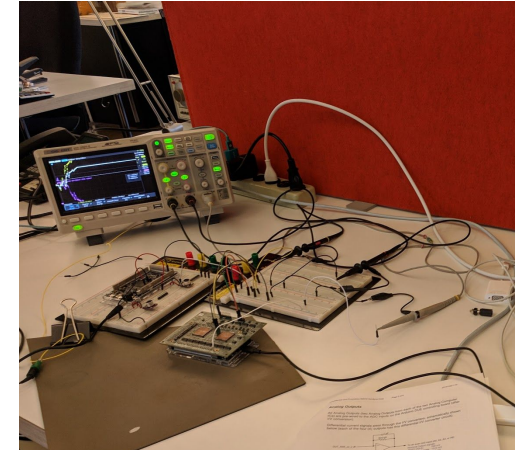
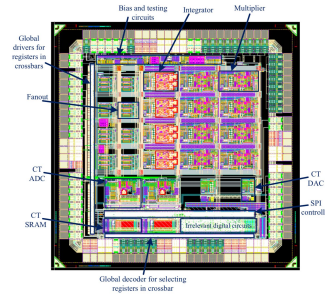
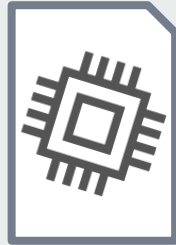
Compiler



Results

Experimental Setup

HCDCv2 Analog Device



COLUMBIA
UNIVERSITY



Guo, Ning, Investigation of Energy-Efficient Hybrid Analog/Digital Approximate Computation in Continuous Time. Columbia University. 2017

Dampened
spring
oscillator

Botulism
neurotoxin
model

Kalman filter

PI controller

Vanderpol
oscillator with
forcing function

Dampened
oscillator

Pendulum

Oscillator

Vanderpol
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Genetic toggle
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Movement of
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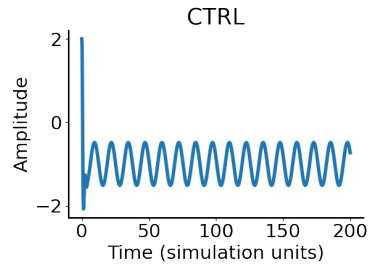
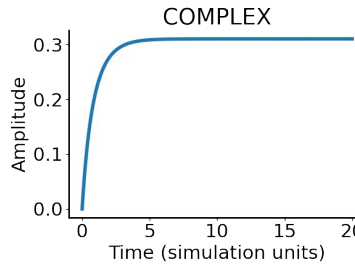
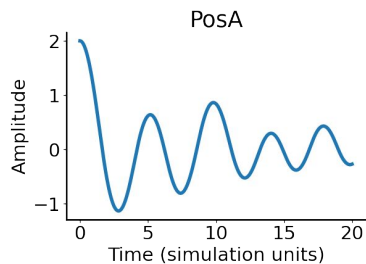
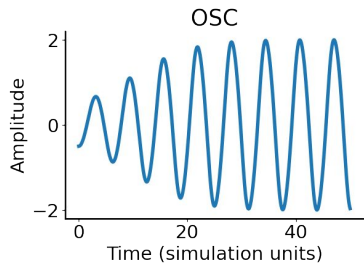
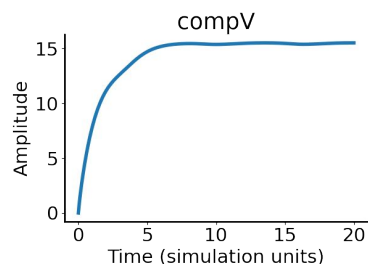
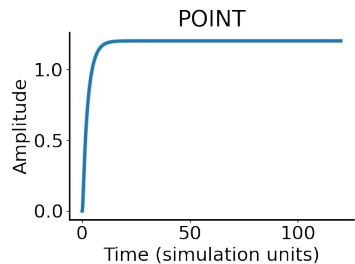
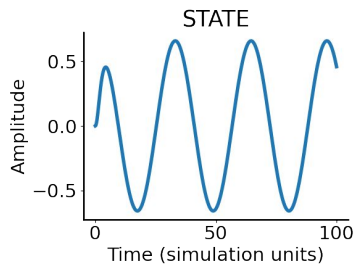
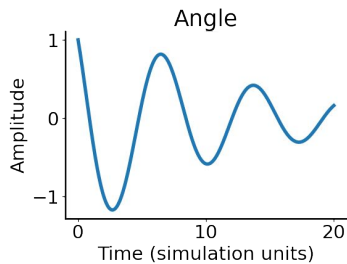
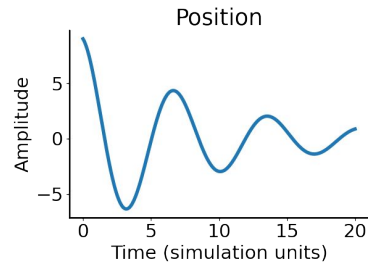
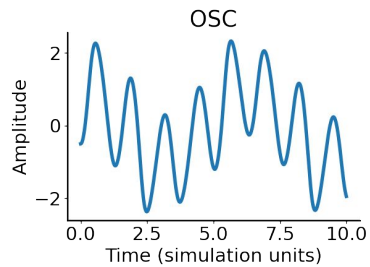
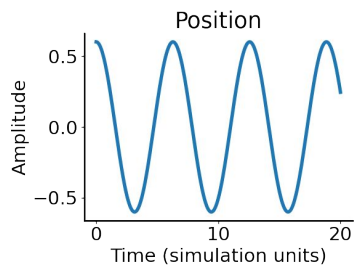
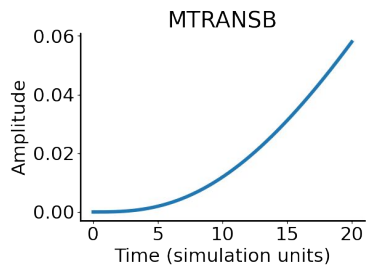
Oscillator

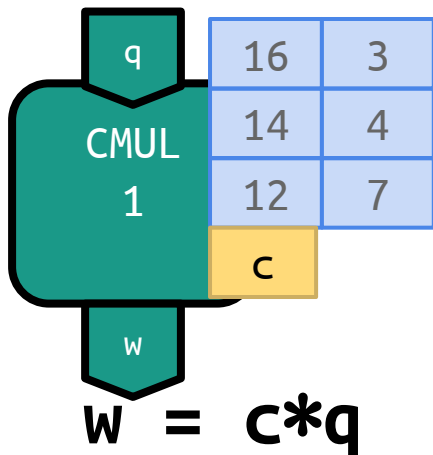
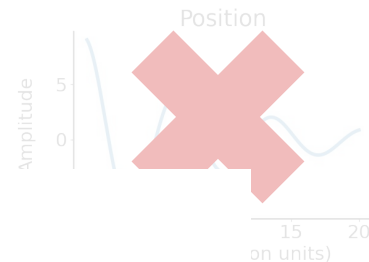
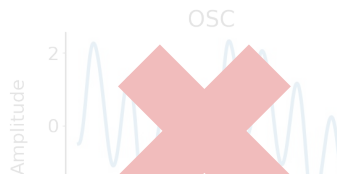
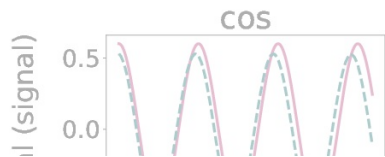
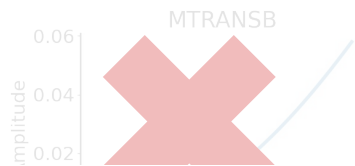
Vanderpol
oscillator

Genetic toggle
switch

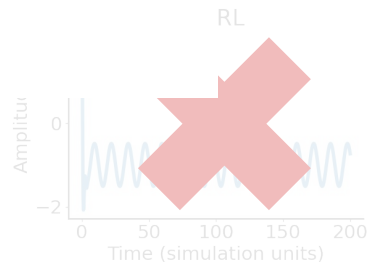
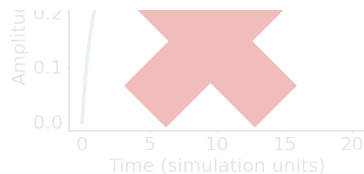
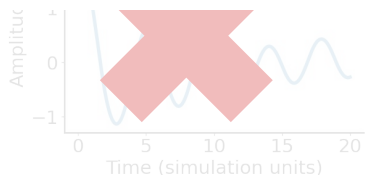
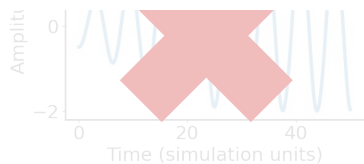
Movement of
heat

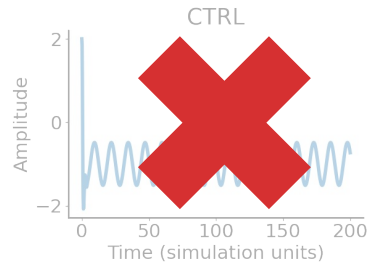
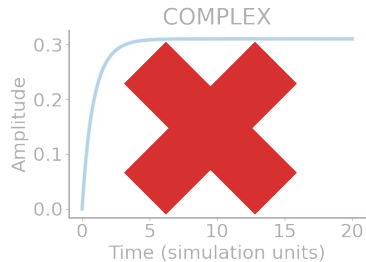
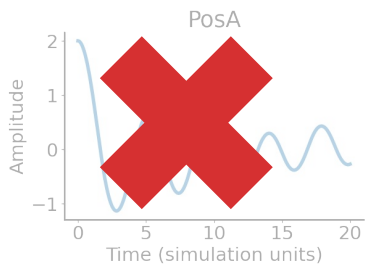
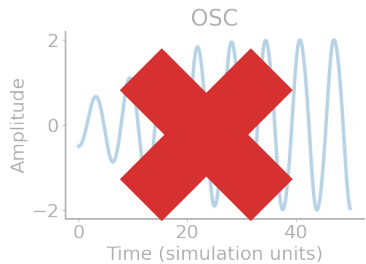
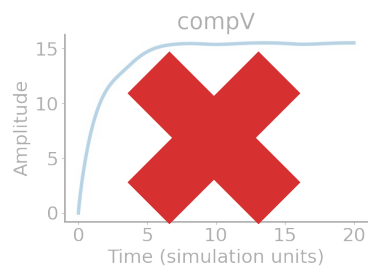
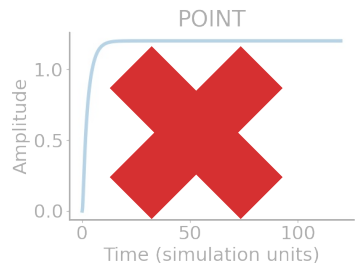
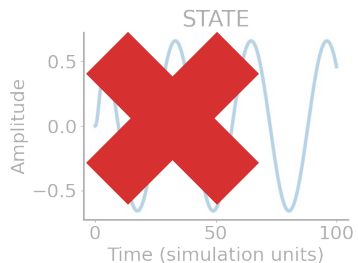
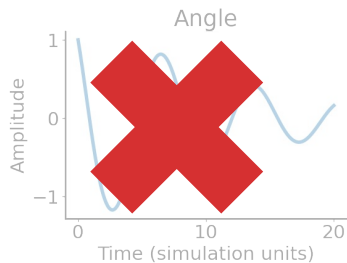
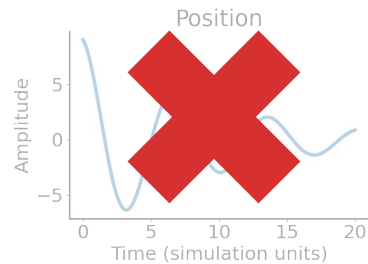
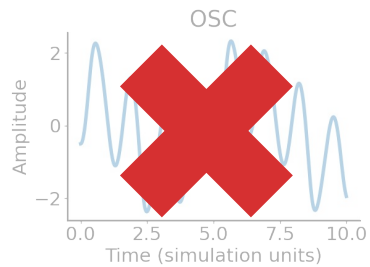
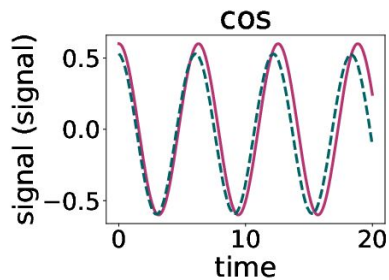
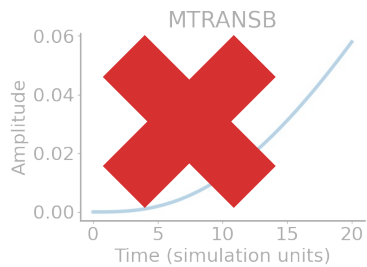
Michaelis
menten
reaction

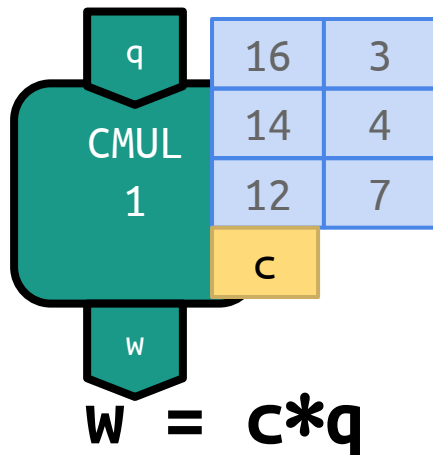
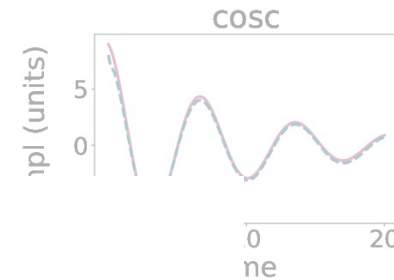
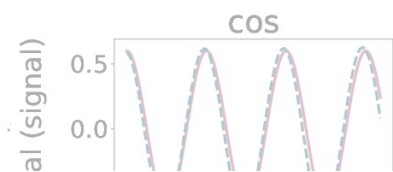
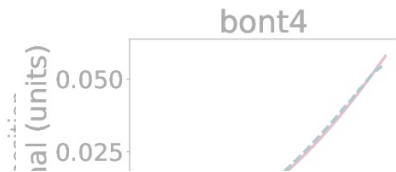




Synthesize configurations without scaling

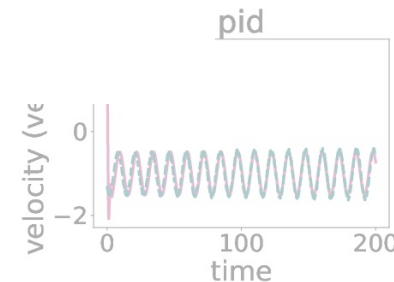
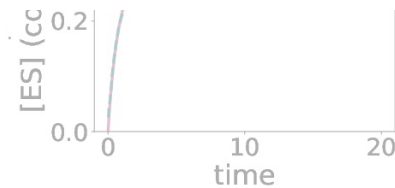
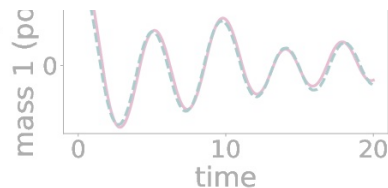
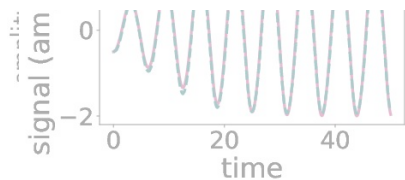


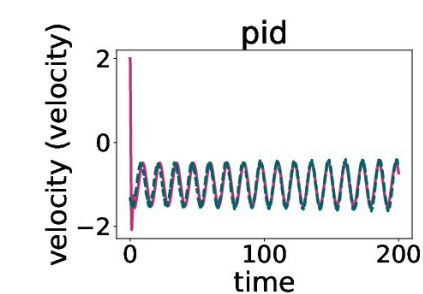
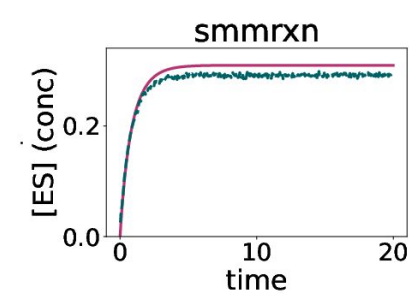
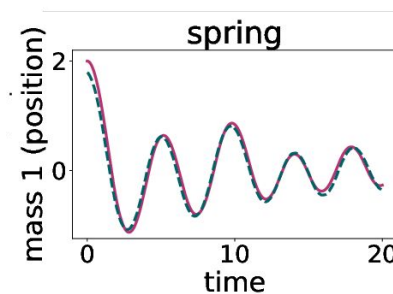
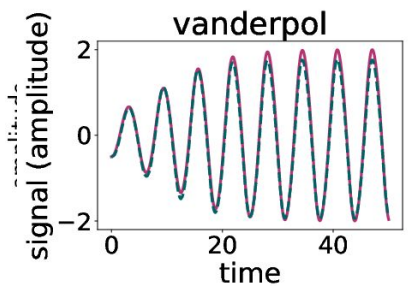
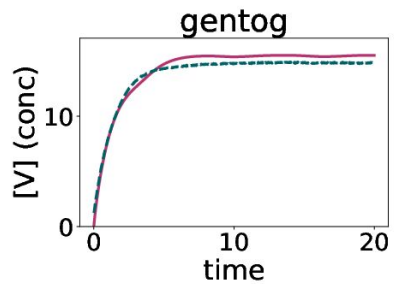
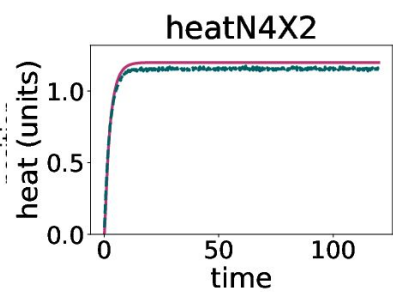
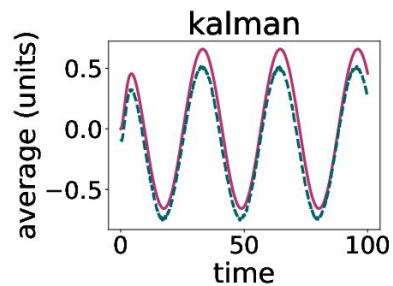
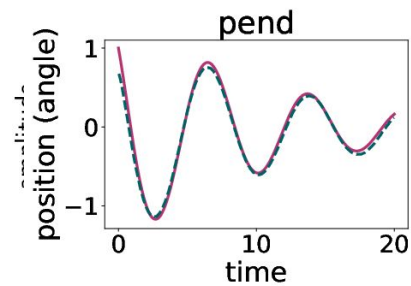
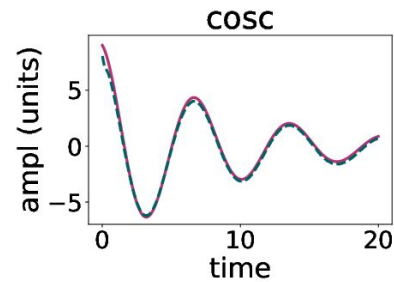
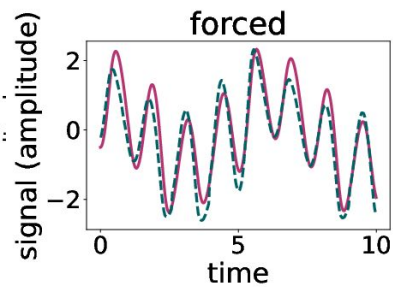
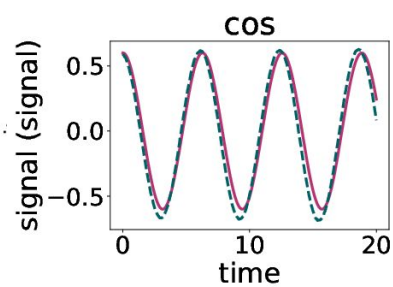
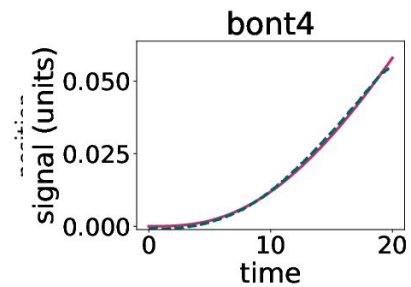


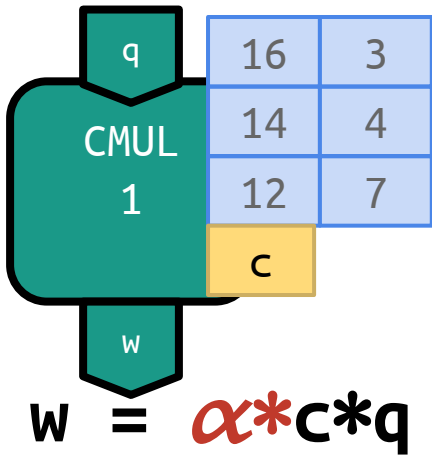
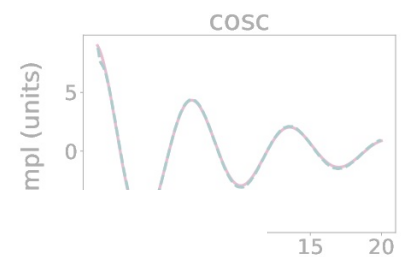
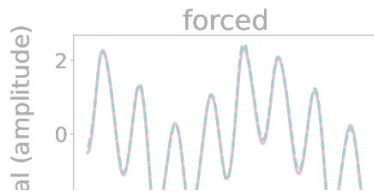
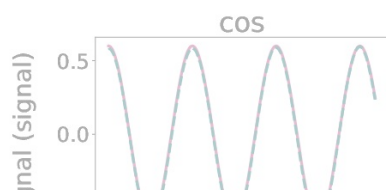


Calibration Objective
Minimize Error

Assume calibrated blocks adhere to specification

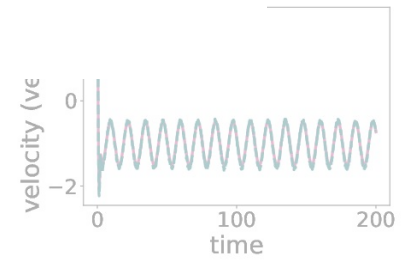
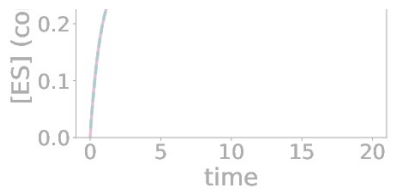
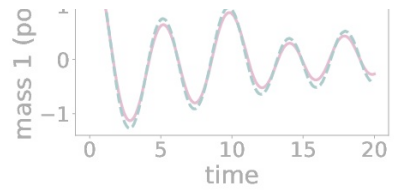
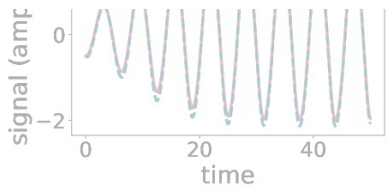


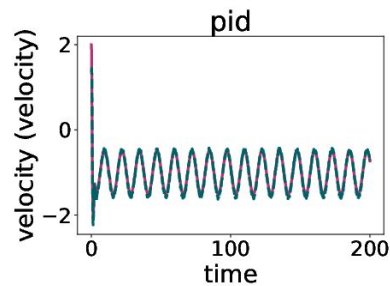
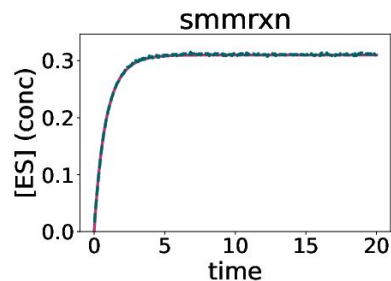
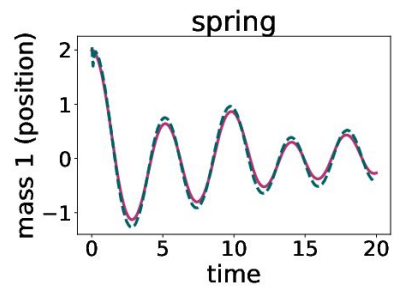
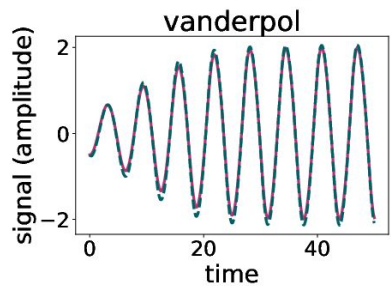
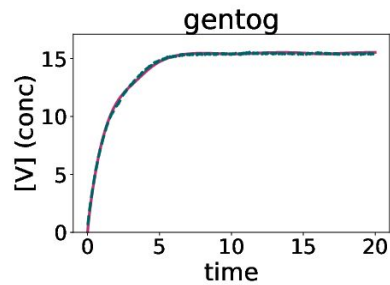
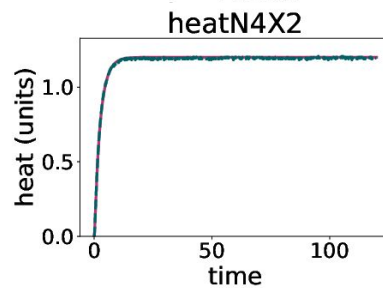
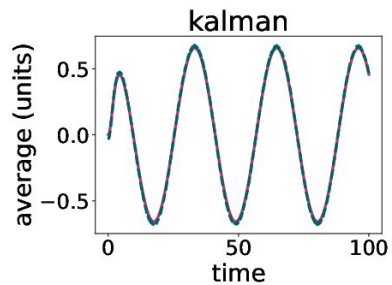
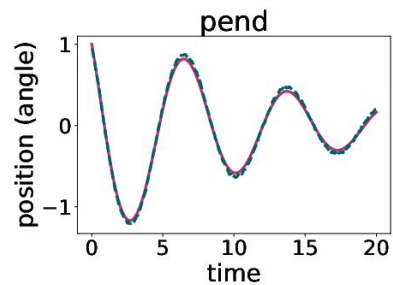
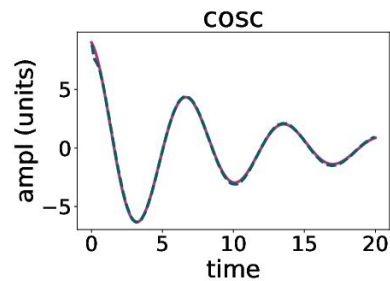
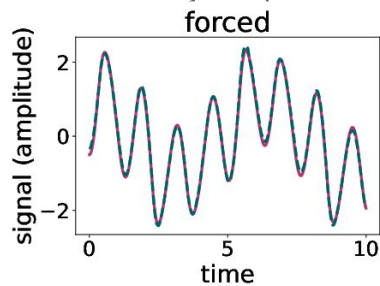
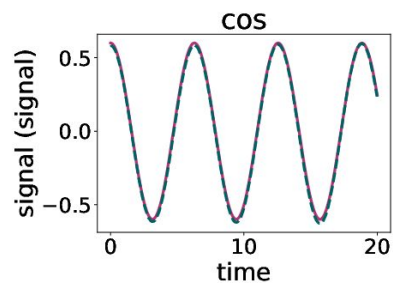
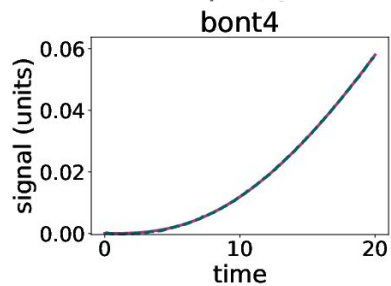




Calibration Objective
Find best fitting model

Allow calibrated blocks to deviate from specification





Dampened
spring
oscillator

Botulism
neurotoxin
model

Kalman filter

PI controller

Vanderpol
oscillator with
forcing function

Dampened
oscillator

Pendulum

Oscillator

Vanderpol
oscillator

Genetic toggle
switch

Movement of
heat

Michaelis
menten
reaction

0.913 mW

0.823 mW

0.864 mW

0.861 mW

0.722 mW

0.395 mW

0.554 mW

0.199 mW

0.849 mW

0.804 mW

0.556 mW

0.526 mW

1.50 ms

0.50 ms

3.32 ms

6.58 ms

3.97 ms

1.34 ms

0.50 ms

1.59 ms

1.25 ms

0.50 ms

9.52 ms

0.52 ms

1.37 μJ

0.41 μJ

2.87 μJ

5.67 μJ

3.37 μJ

0.53 μJ

0.28 μJ

0.32 μJ

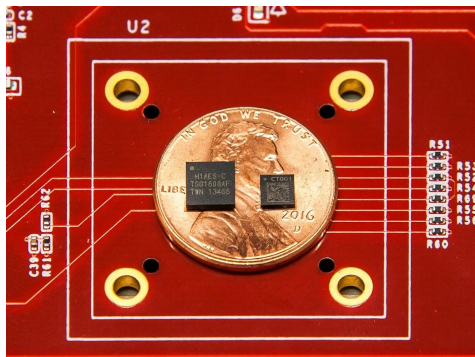
0.90 μJ

0.40 μJ

5.30 μJ

0.28 μJ

Future Directions



Secure Co-Processors



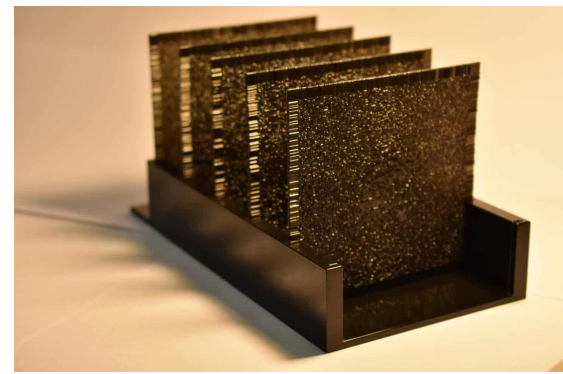
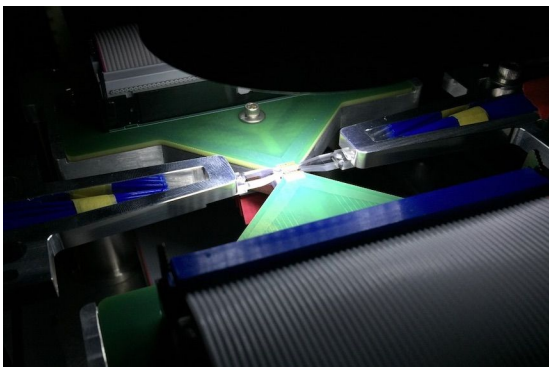
DSPs



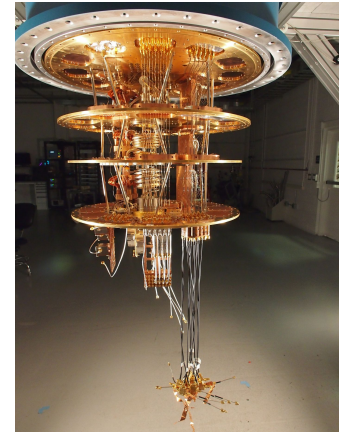
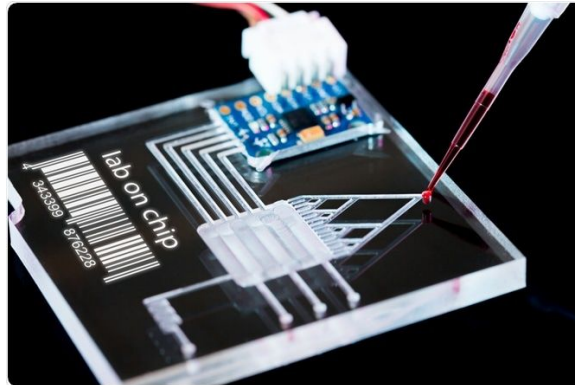
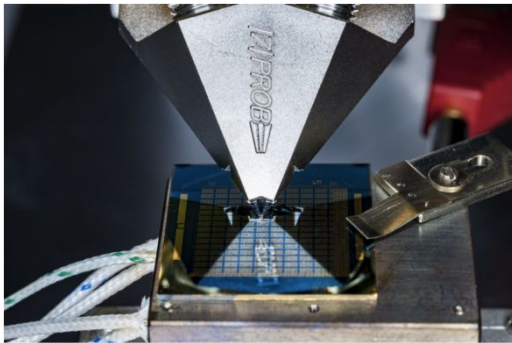
TPUs

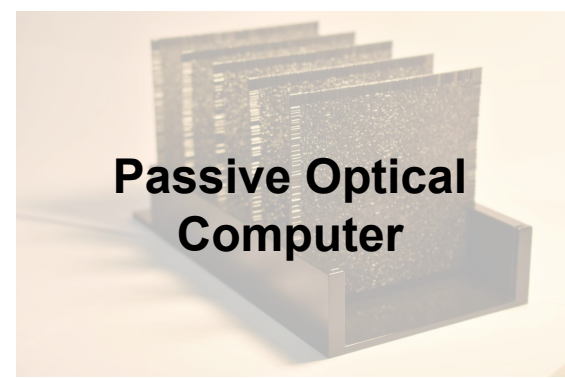
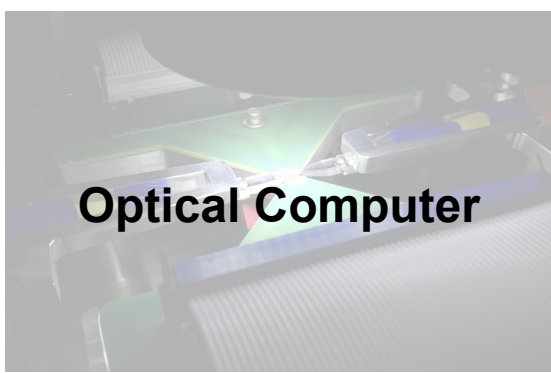


GPUs

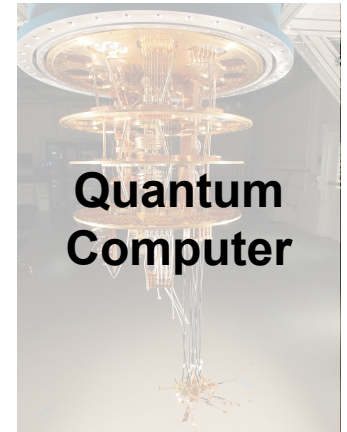
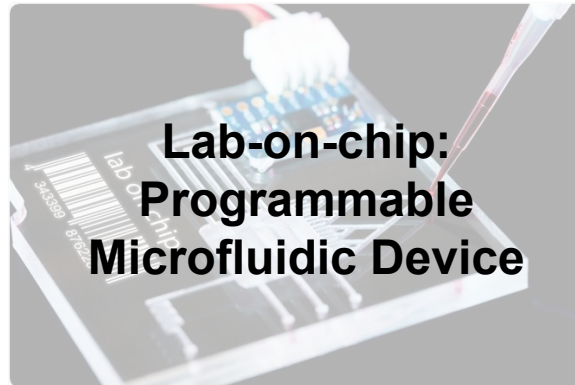
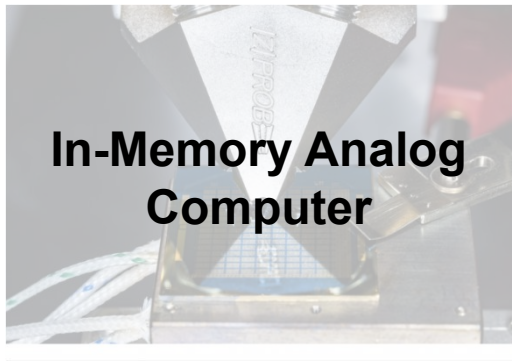


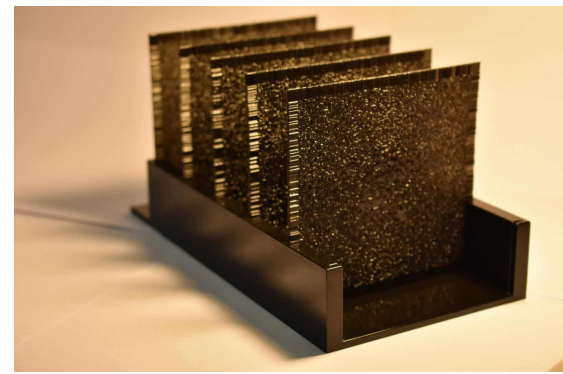
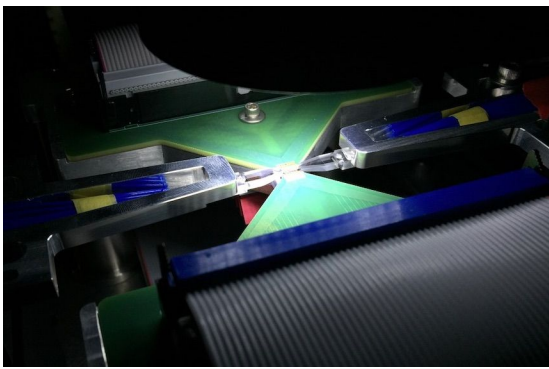
**Many exciting special-purpose hardware platforms
on the horizon**



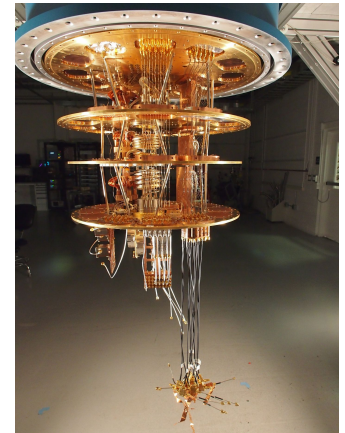
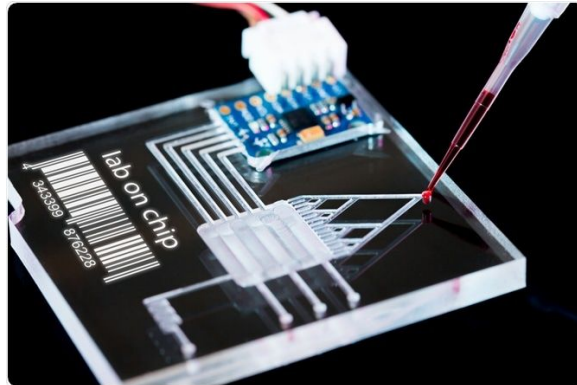
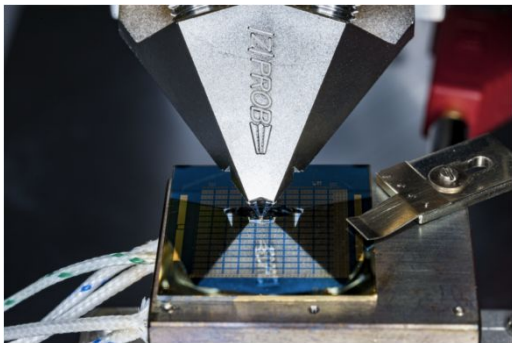


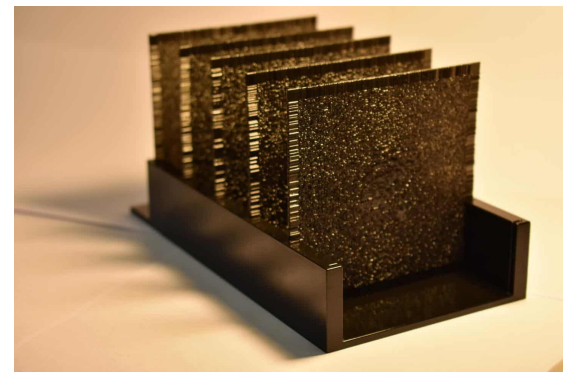
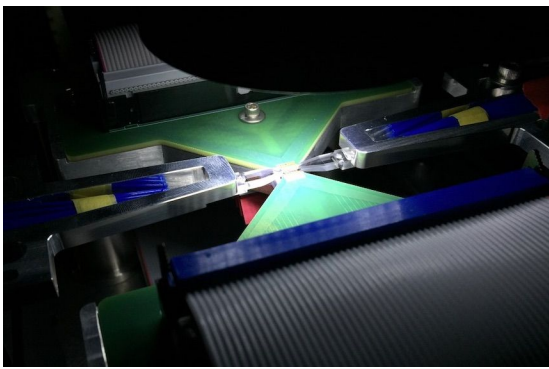
**Many exciting special-purpose hardware platforms
on the horizon**



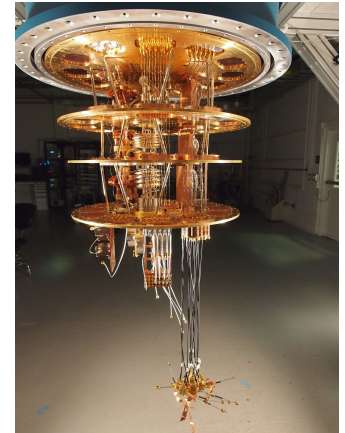
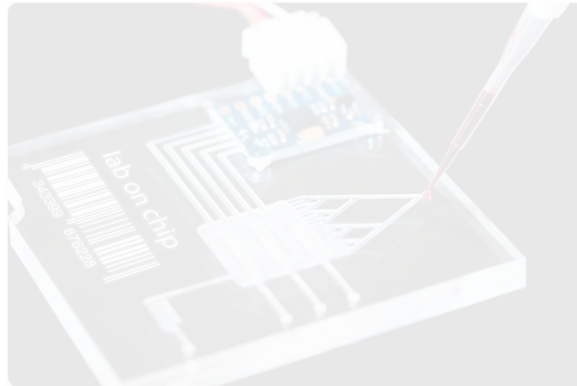
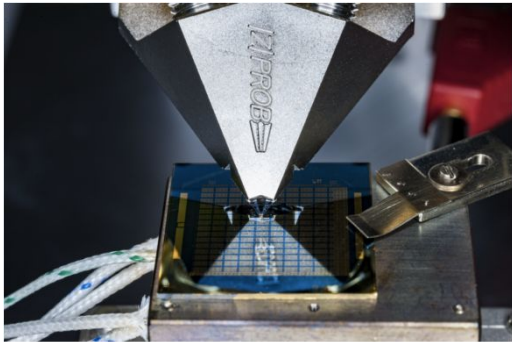


Common Theme: **Analog Behavior!**



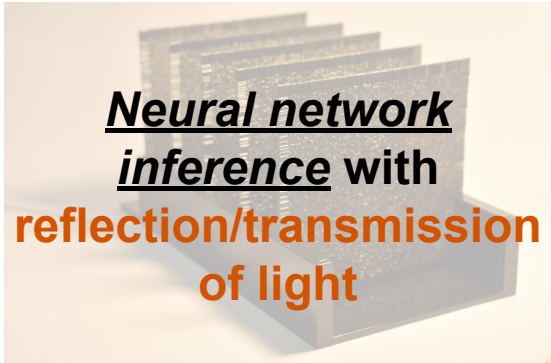


Some devices **leverage analog behavior** to perform computation



A photograph of an optical setup for multiplication using interference. It shows a green laser beam passing through a series of optical components, including waveguides and phase shifters, on a substrate.

**Multiplication with
optical interference**

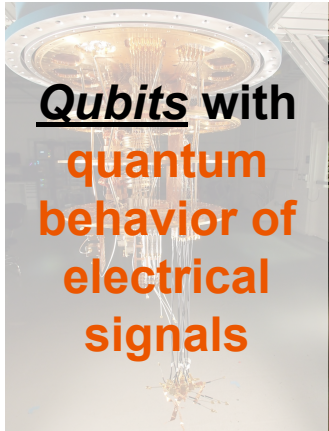
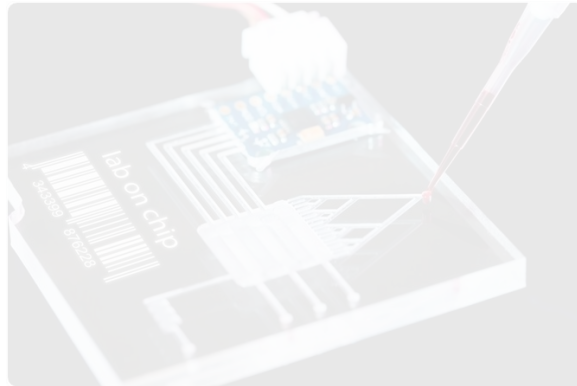
A photograph of a neural network inference setup using reflection and transmission of light. It shows a series of stacked, rectangular optical components, possibly waveguides or photonic chips, arranged in a row.

**Neural network
inference with
reflection/transmission
of light**

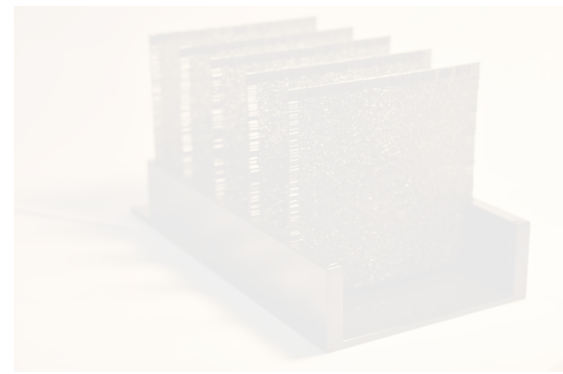
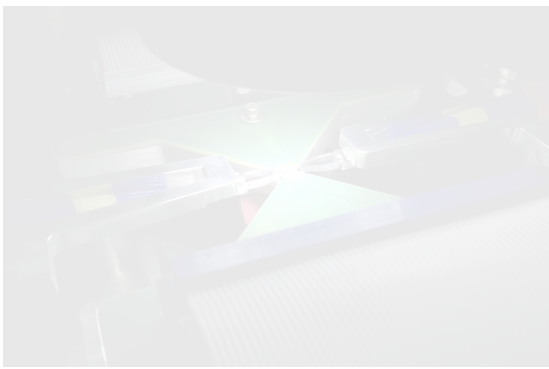
Some devices **leverage analog behavior** to efficiently perform computation

A photograph of a multiply-accumulate setup. It shows a complex, multi-layered structure, possibly a photonic chip, with various components and connections. The text "17-proc" is visible on the structure.

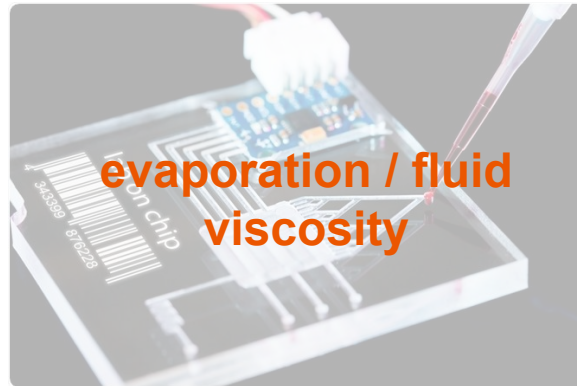
**Multiply-accumulate
with **ohm's/kirchoff's
law****

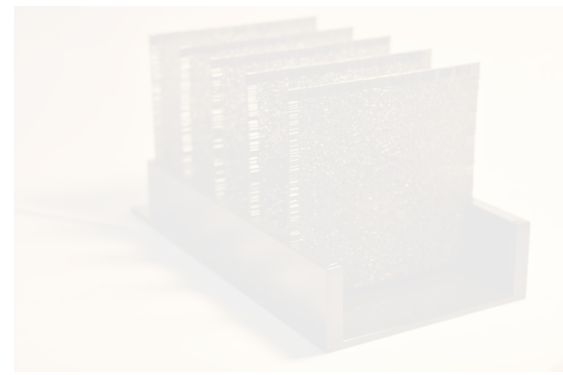
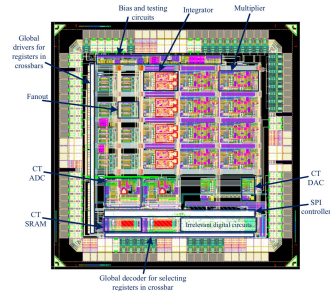
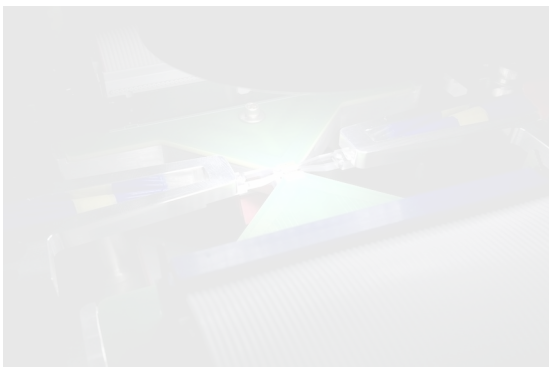
A photograph of qubits with quantum behavior of electrical signals. It shows a complex, multi-layered structure, possibly a photonic chip, with various components and connections. The text "Qubits with quantum behavior of electrical signals" is visible on the structure.

**Qubits with
quantum
behavior of
electrical
signals**

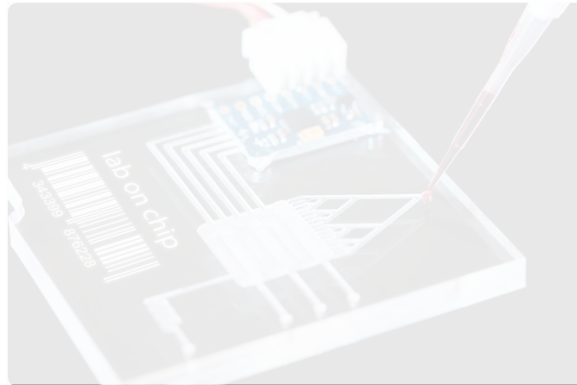


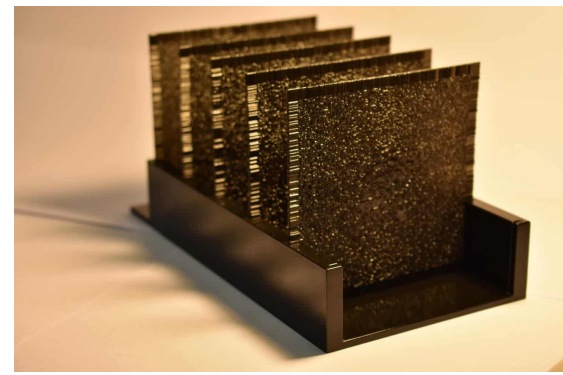
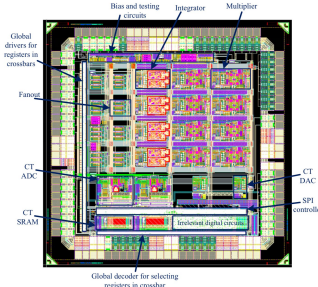
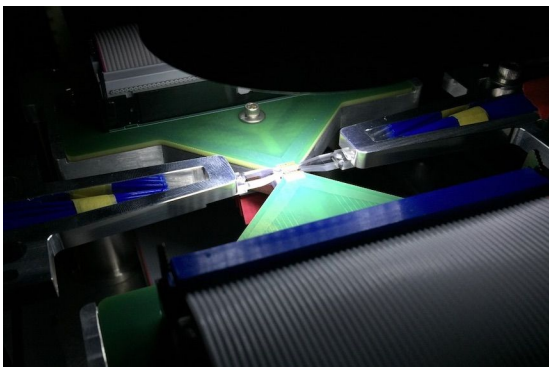
**Some computational models/technologies are
adversely affected by analog behavior**



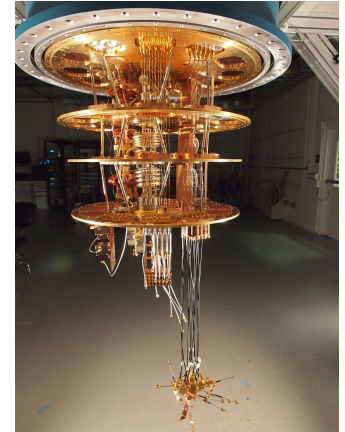
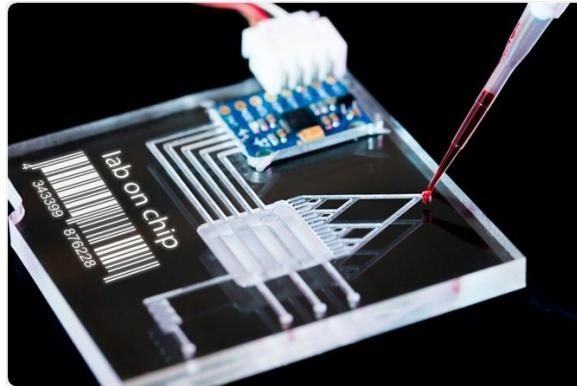
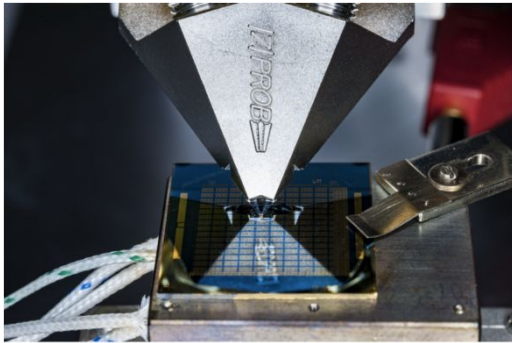


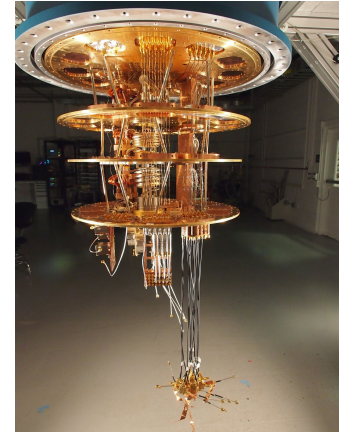
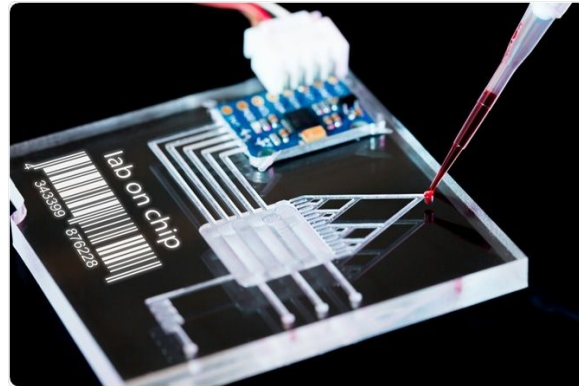
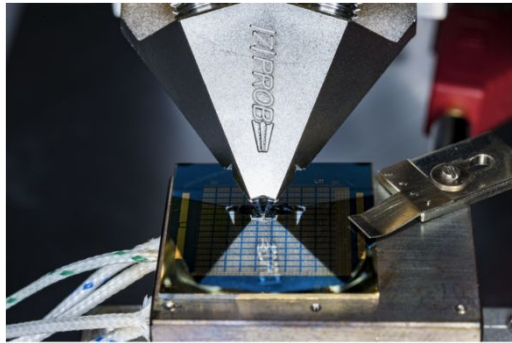
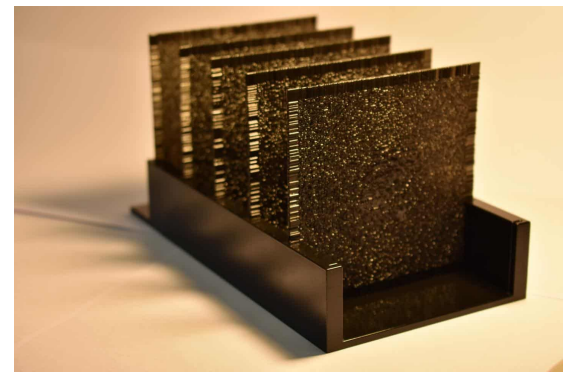
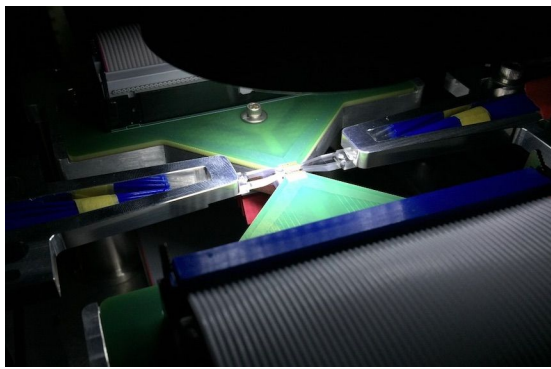
Some devices both **leverage** and are **adversely affected by** analog behavior





Some devices both leverage and are adversely affected by analog behavior







Contributions

- New compilation techniques for reconfigurable analog devices
 - Circuit synthesis techniques
 - Automated scaling techniques
- First-ever compilation toolchain for a real-world analog device of this class
- Experimental results evaluating effectiveness of compilation toolchain



Backup

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, \dots, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

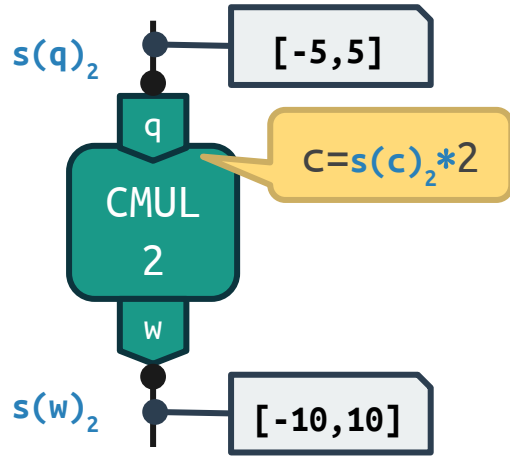
Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

Operating Range Constraints



$$q, w \in [-2, 2] \mu A$$
$$c \in [-1, 1]$$

$$s(q)_2 * [-5, 5] \subseteq [-2, 2] \mu A$$
$$s(w)_2 * [-10, 10] \subseteq [-2, 2] \mu A$$
$$s(c)_2 * 2 \subseteq [-1, 1]$$

Automatic Scaling



Convex Optimization

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Variables: $\tau, s(x)_2, s(z)_2, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

Execution Speed Constraints

Quality Constraints

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Factor Constraints

Automatic Scaling

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Connection Constraints

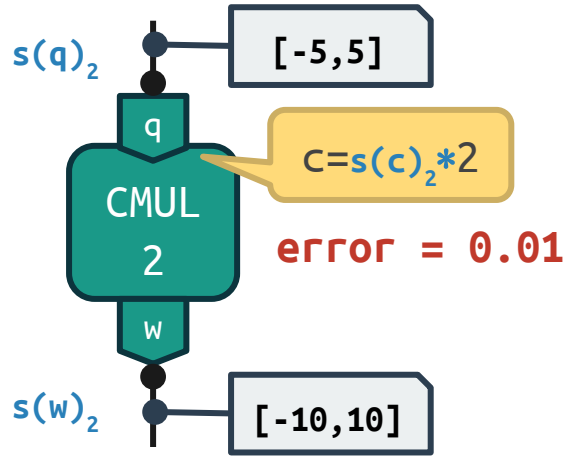
Factor Constraints

minimum analog and digital SNR

Digital
SNR

Analog
SNR

Quality Constraints



$w \sim \text{Normal}(w, 0.02 \mu\text{A})$

$$s(w)_2 * \|[-10, 10] \| (0.02 \mu\text{A})^{-1} \geq$$

Analog
SNR

$$s(c)_2 * 2 (0.01)^{-1} \geq$$

Digital
SNR

Automatic Scaling



Convex Optimization

Formulated as geometric programming problem.

Variables: $\tau, s(x)_2, s(z)_2, s(c)$

Objective: Maximize execution speed

Minimize τ

Subject to:

Operating Range Constraints

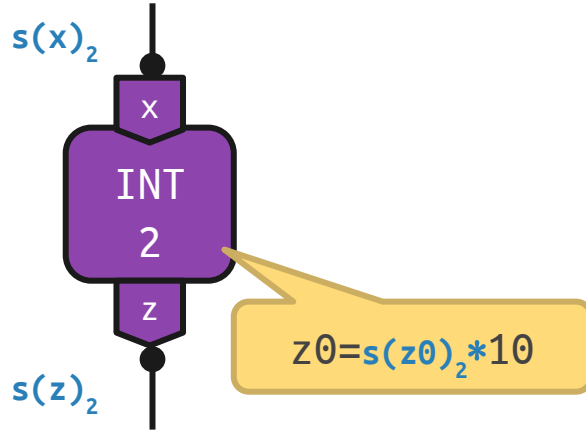
Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

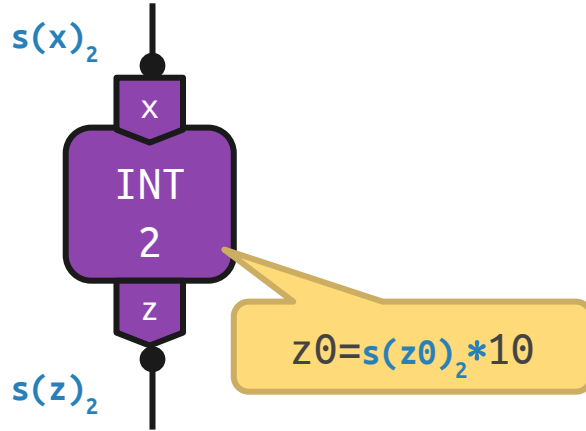
Factor Constraints



$$s(z)_2 * z = \int 0.93 * 0.5 * (s(x)_2 * x) \tau^{-1} dt_{hw}$$

$$s(z)_2 * z(0) = 0.77 * s(z_0)_2 * 10$$

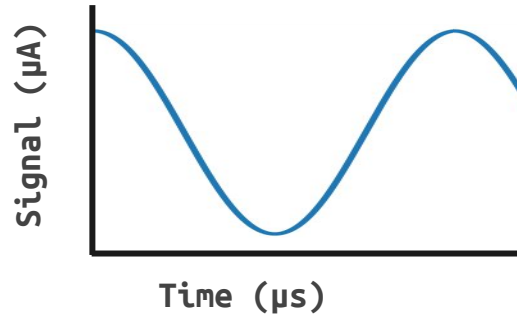
Factor Constraints



$$s(z)_2 * z = \int 0.93 * 0.5 * (s(x)_2 * x) \tau^{-1} d\tau_{hw}$$

$$s(z)_2 * z(0) = 0.77 * s(z_0)_2 * 10$$

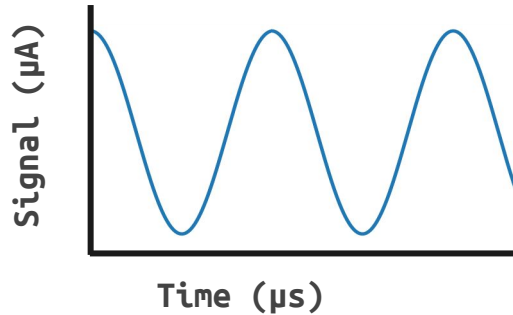
Time Scaling Differential Equations



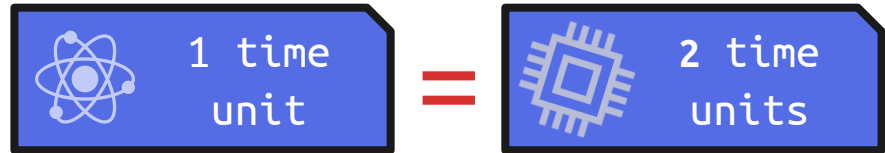
$$\tau = 2.0$$

$$p = \int v \tau^{-1} dt_{hw}$$

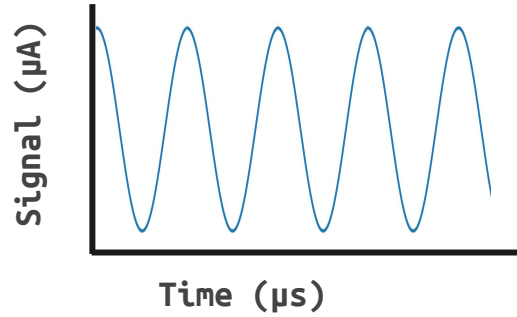
$$v = \int -0.25 * p \tau^{-1} dt_{hw}$$



$$\tau = 1.0$$



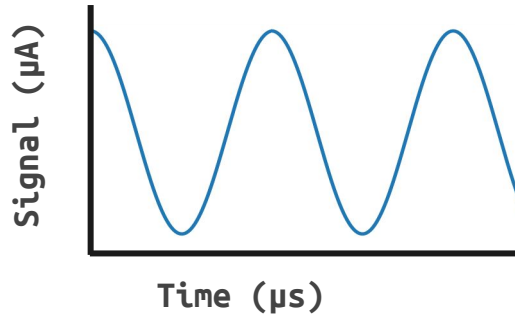
Time Scaling Differential Equations



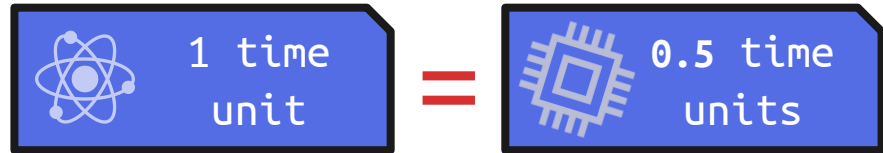
$$\tau = 0.5$$

$$p = \int v \tau^{-1} dt_{hw}$$

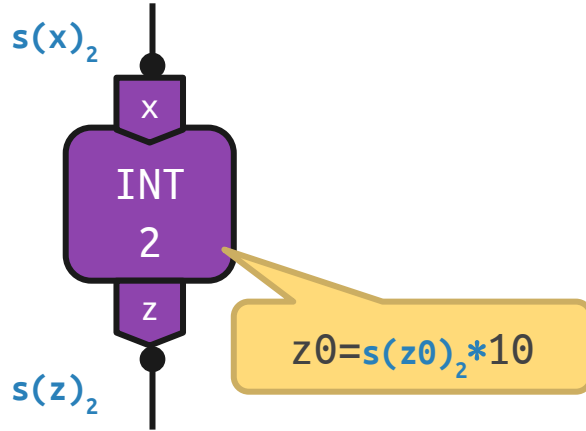
$$v = \int -0.25 * p \tau^{-1} dt_{hw}$$



$$\tau = 1.0$$



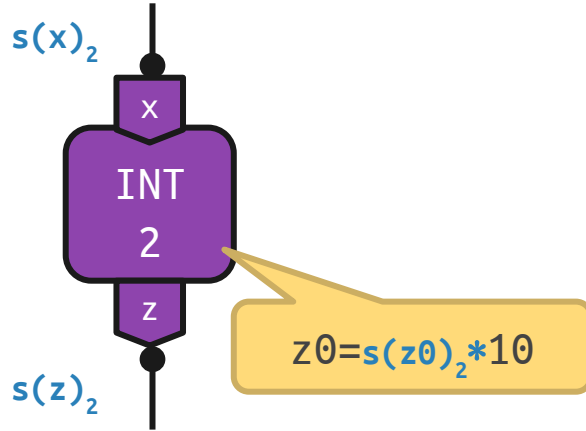
Factor Constraints



$$s(z)_2 * z = \int 0.93 * 0.5 * (s(x)_2 * x) \tau^{-1} dt_{hw}$$

$$s(z)_2 * z(0) = 0.77 * s(z_0)_2 * 10$$

Factor Constraints

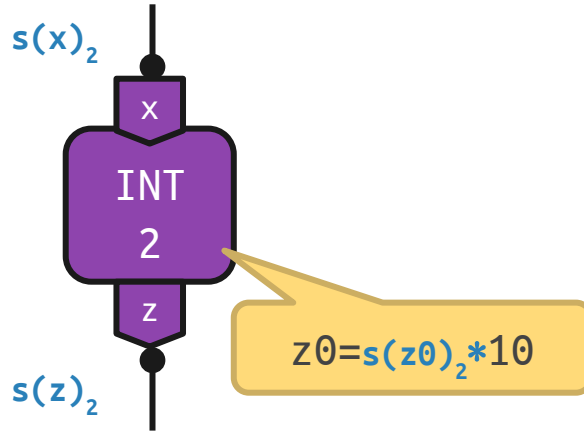


$$s(z)_2 * z = \int 0.93 * 0.5 * (s(x)_2 * x) \tau^{-1} dt_{hw}$$

$$s(z)_2 * z(0) = 0.77 * s(z0)_2 * 10$$

Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)

Factor Constraints

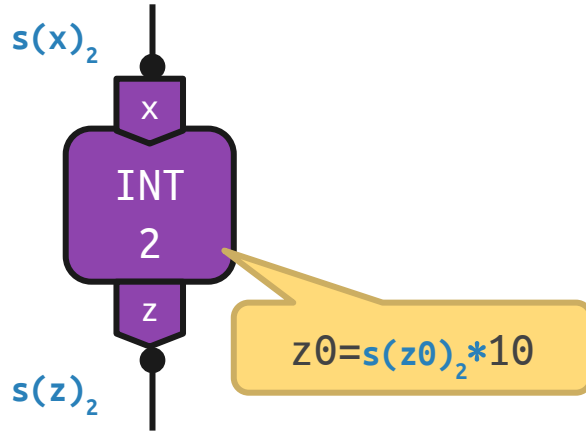


$$s(z)_2 * z = (0.93 * s(x)_2 * \tau^{-1}) \int 0.5 * (x) dt_{hw}$$

$$s(z)_2 * z(0) = (0.77 * s(z0)_2) 10$$

Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)

Factor Constraints



$$z = \int 0.5 * (x) dt_{hw}$$

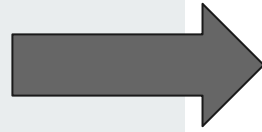
$$s(z)_2 = 0.93 * s(x)_2 * \tau^{-1}$$

$$z(0) = 10$$

$$s(z)_2 = 0.77 * s(z0)_2$$

Goal: Factor out and eliminate **scaling factors** and **manufacturing variations** from dynamics (black)

Unscaled



Algebraically
Equivalent to



Convex Optimization Problem

Minimize τ

Subject to:

Operating Range Constraints

Execution Speed Constraints

Quality Constraints

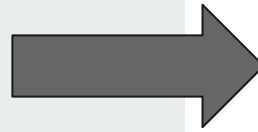
Connection Constraints

Factor Constraints

Unscaled



Algebraically
Equivalent to



Convex Optimization Problem

Minimize τ

Subject to:

Operating Range Constraints

Execution Speed Constraints

Quality Constraints

Connection Constraints

Factor Constraints

Unscaled



Algebraically
Equivalent to



Solved using Convex Solver

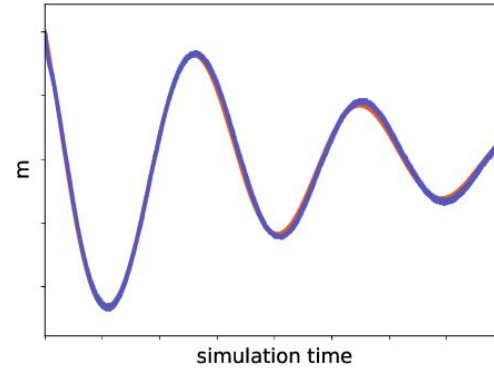
What can go wrong?

Challenges

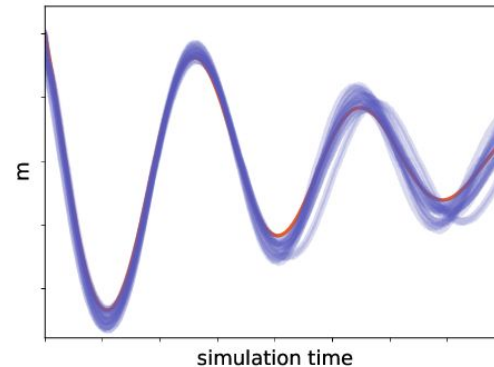


Not all generated configurations deliver good results

Best Pendulum Execution



All Pendulum Executions



Challenges



Not all generated configurations deliver good results

Some configurations are actually better than others

Design method for ranking device configurations

Challenges



Not all generated configurations deliver good results

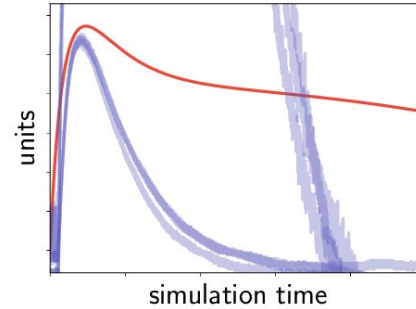
Not all manufactured block behaviors captured by compiler

Profile and statically compensate for other block behaviors.

Challenges

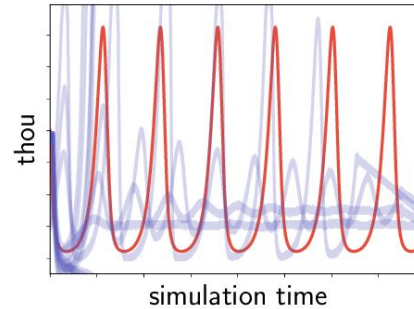
Not all dynamical systems can be accurately executed on device

Dynamic Range too High



Erythropoietin receptor model

System is Sensitive to Error



Predator-Prey simulation

Challenges



**Not all dynamical
systems can be
accurately
executed on device**

**Automatically
characterize
dynamical systems**

Challenges



Not all dynamical systems can be accurately executed on device

Automatically characterize dynamical systems

How much error can the system tolerate?

Challenges

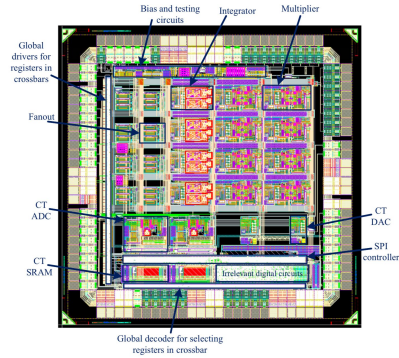
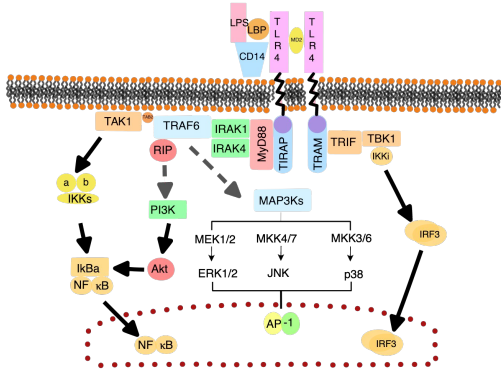
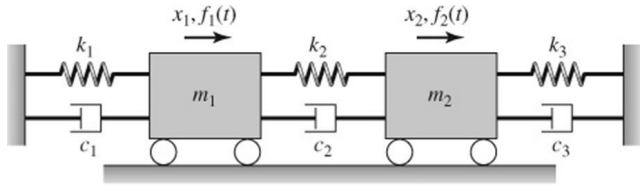


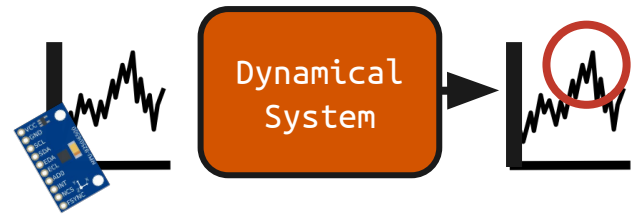
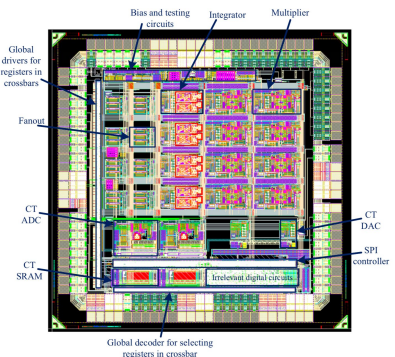
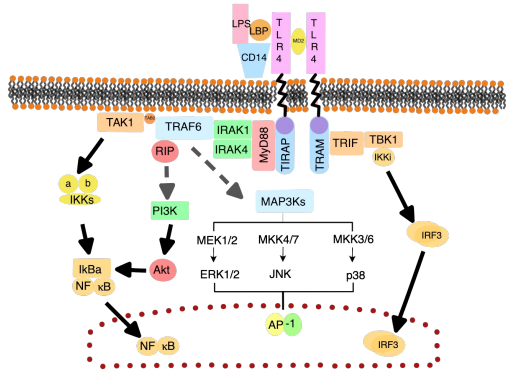
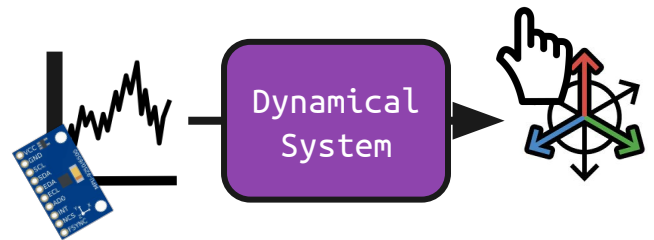
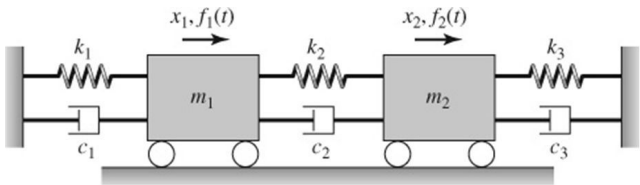
Not all dynamical systems can be accurately executed on device

Automatically characterize dynamical systems.

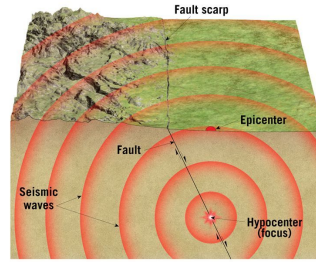
How much error can the system tolerate?

Where in the computation does it tolerate error?

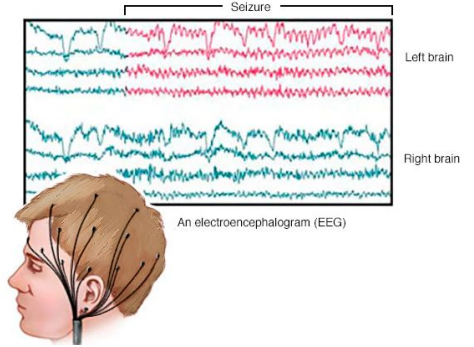




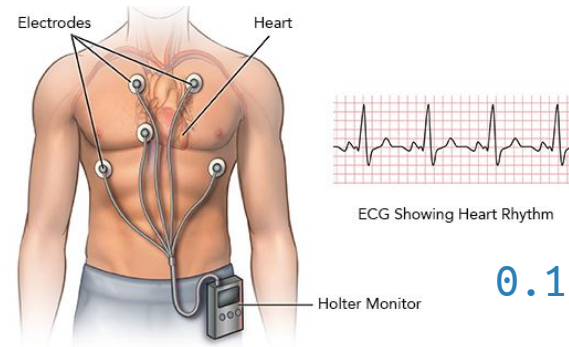
Many interesting signals are slow-evolving



0.2 Hz to 20 Hz



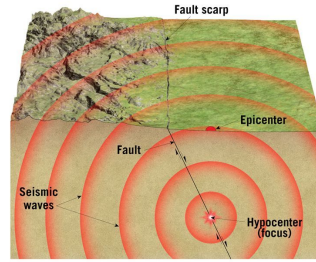
1 Hz-100 Hz



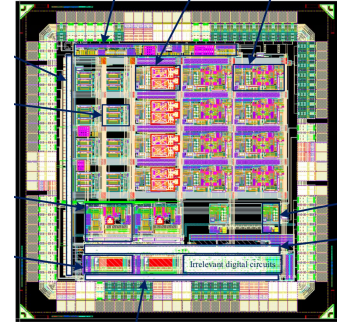
0.1 Hz-1 kHz

Many interesting
signals are
slow-evolving

Slow signals are
difficult to work
with in pure analog



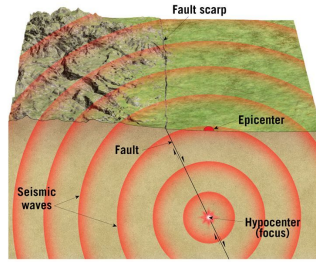
0.2 Hz to 20 Hz



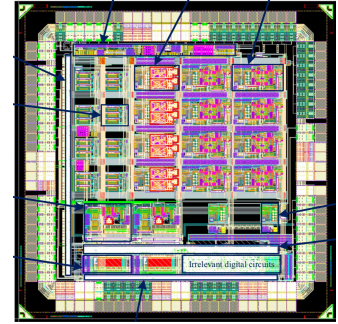
126,000 Hz

Many interesting signals are slow-evolving

Slow signals are difficult to work with in pure analog



0.2 Hz to 20 Hz

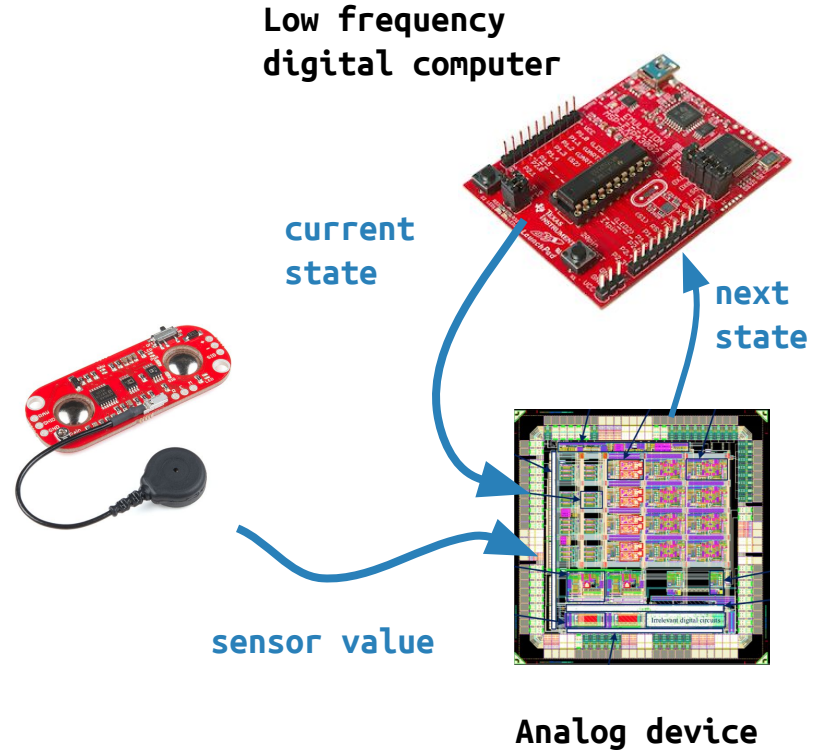


126,000 Hz

Dramatically slow down dynamical system to process signal

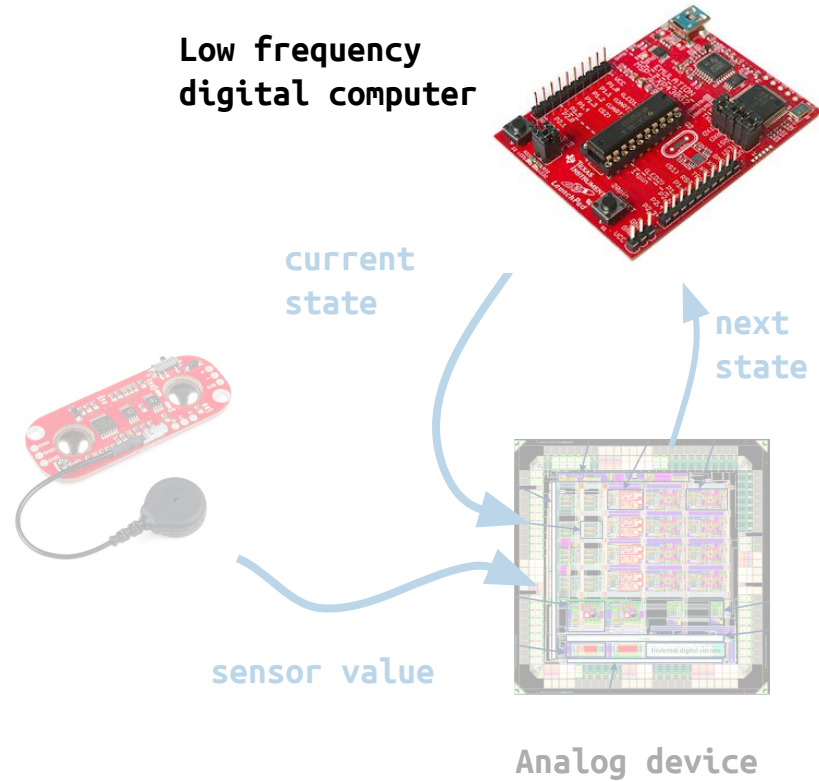
Sensitive to noise and error

Hybrid digital-analog computation



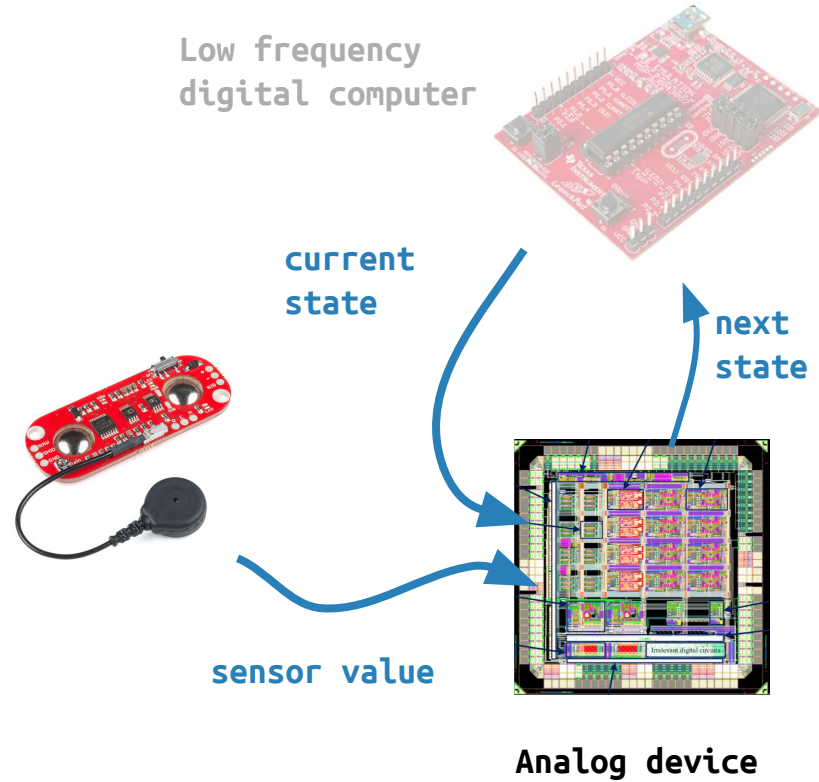
Hybrid digital-analog computation

Digitally store long-running state



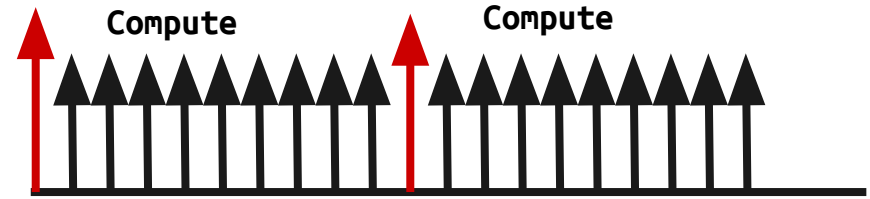
Hybrid digital-analog computation

Process external input and update state in analog



Hybrid digital-analog computation

Pure Digital



Hybrid

