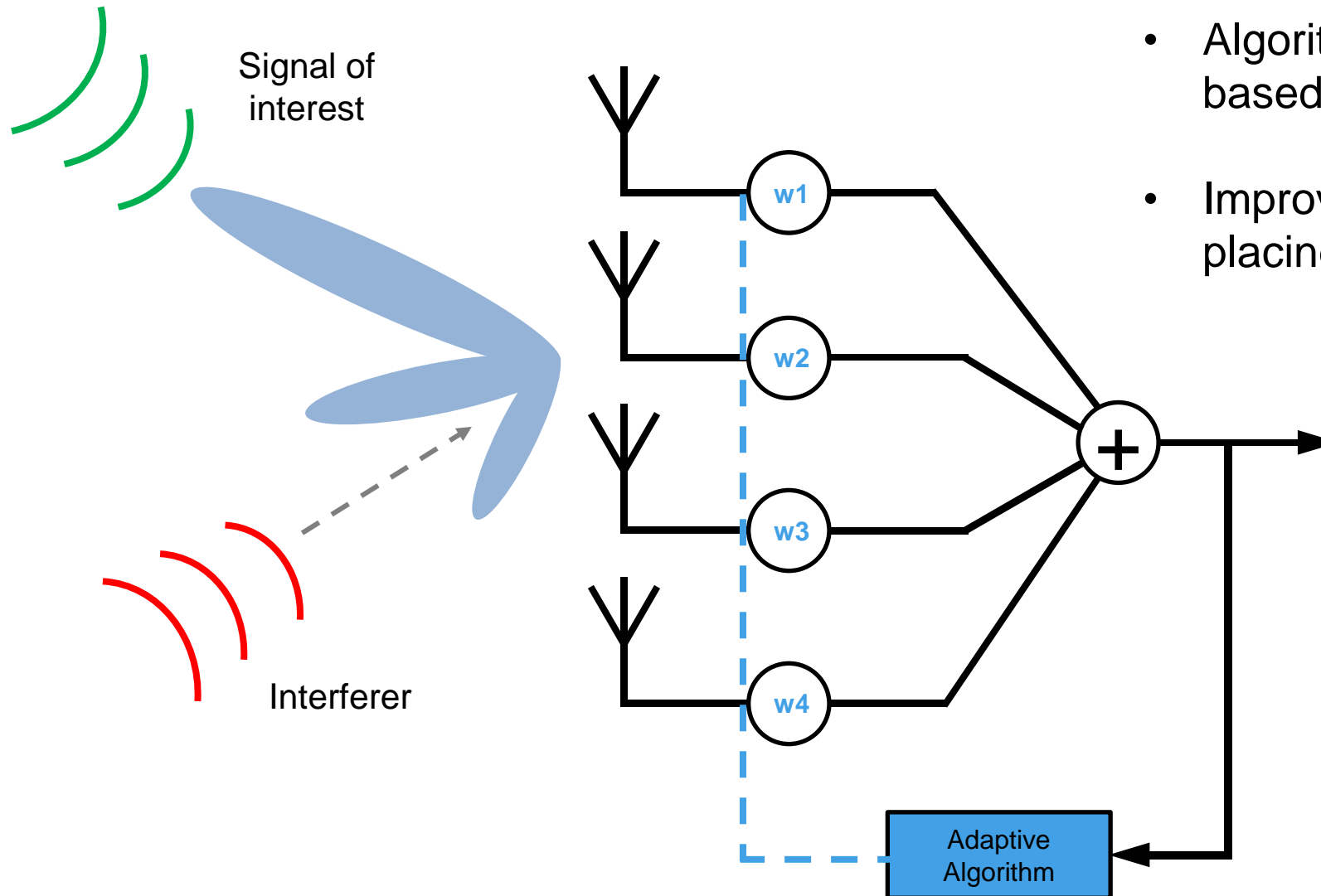


FPGA Adaptive Beamforming with HDL Coder and Zynq RFSoC

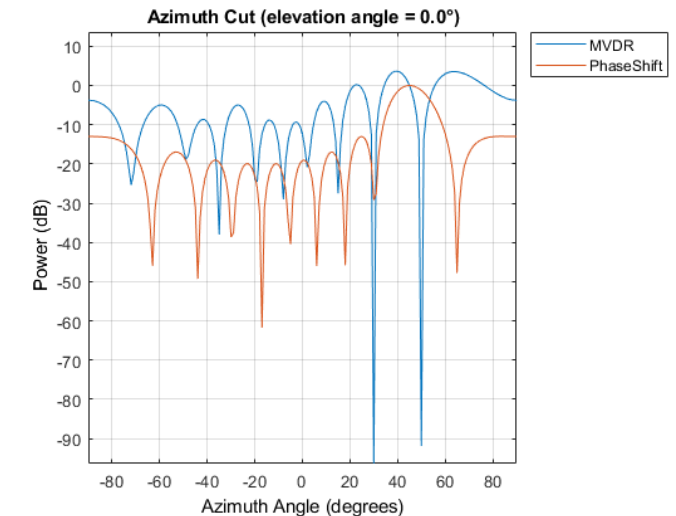
Agenda

- Introduction: Motivation and Challenges
 - Applications: Radar, Comms and Wireless
 - Hardware FPGA challenges
- Theory and Implementation
 - Linear algebra
 - QR Decomposition
 - Matrix Divide
- Zynq RFSoc and HDL Coder Implementation
 - MATLAB MVDR reference code
 - HDL Coder implementation
 - Hardware Prototyping – live demo

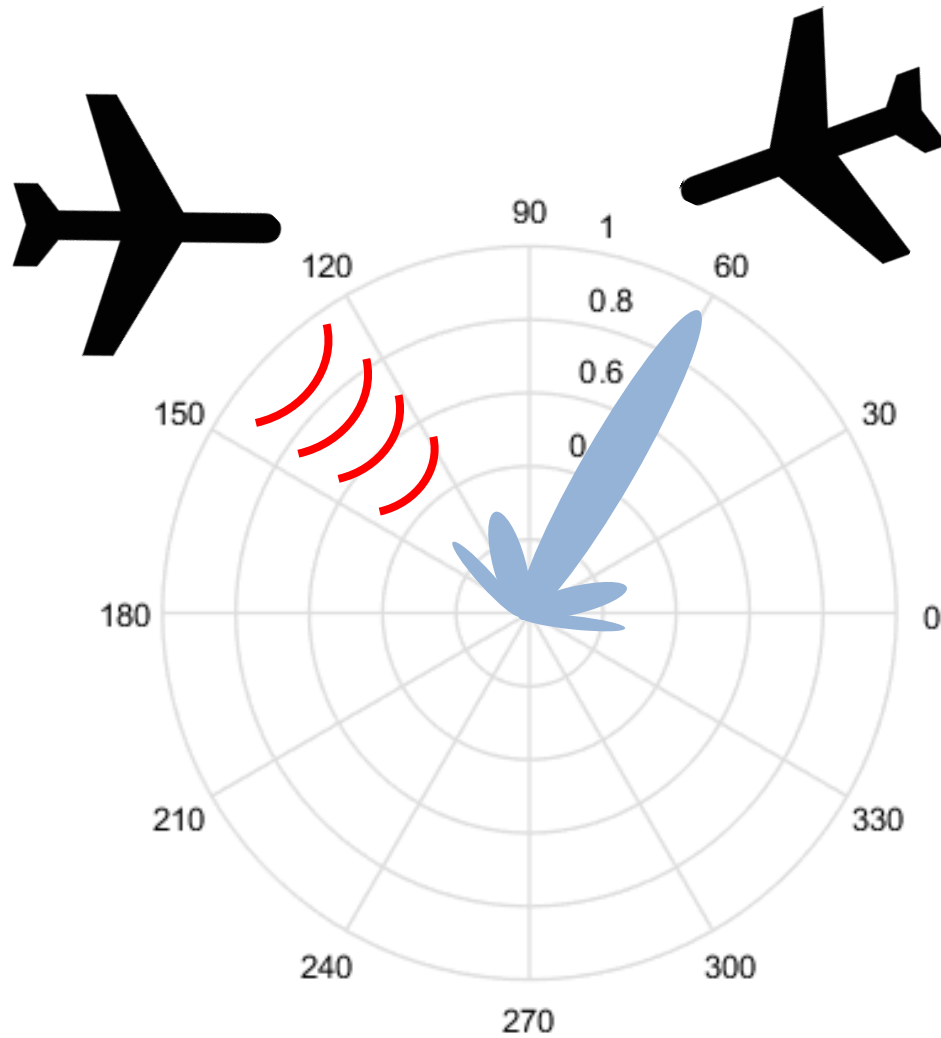
Adaptive Beamforming



- Algorithm chooses optimal weights based on receive data statistics
- Improve SNR by automatically placing nulls at interference angles



Applications: Radar

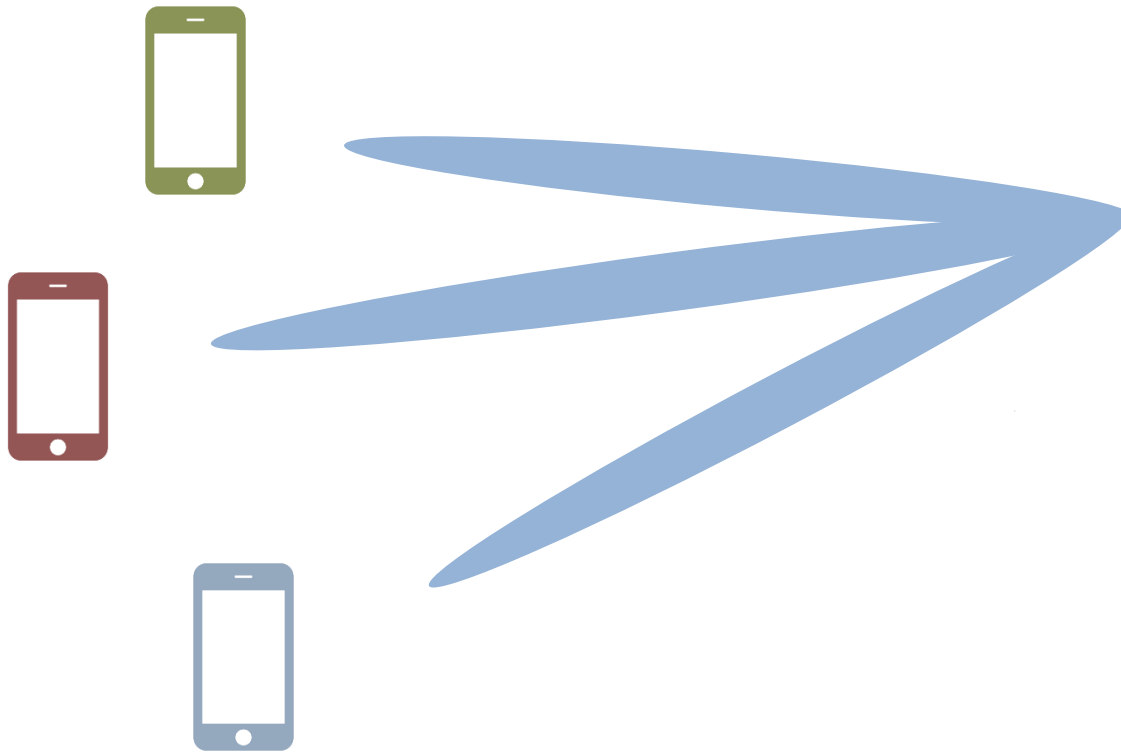


- Increase angular resolution
- Suppress interference



Applications: 5G

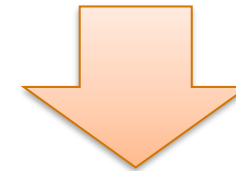
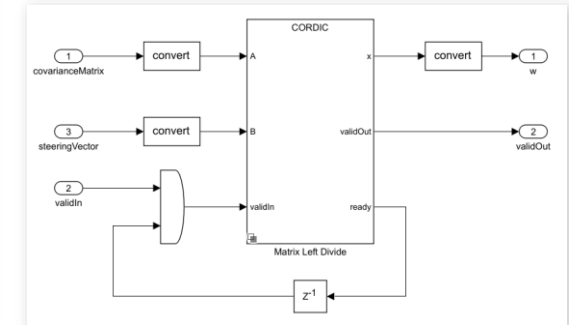
- Increase number of simultaneous users
- Improve throughput and coverage



FPGA Implementation Challenges

- Fixed-Point Math
- Performance vs Area tradeoffs
- Data Rate vs Clock Rate
- Project Timeline

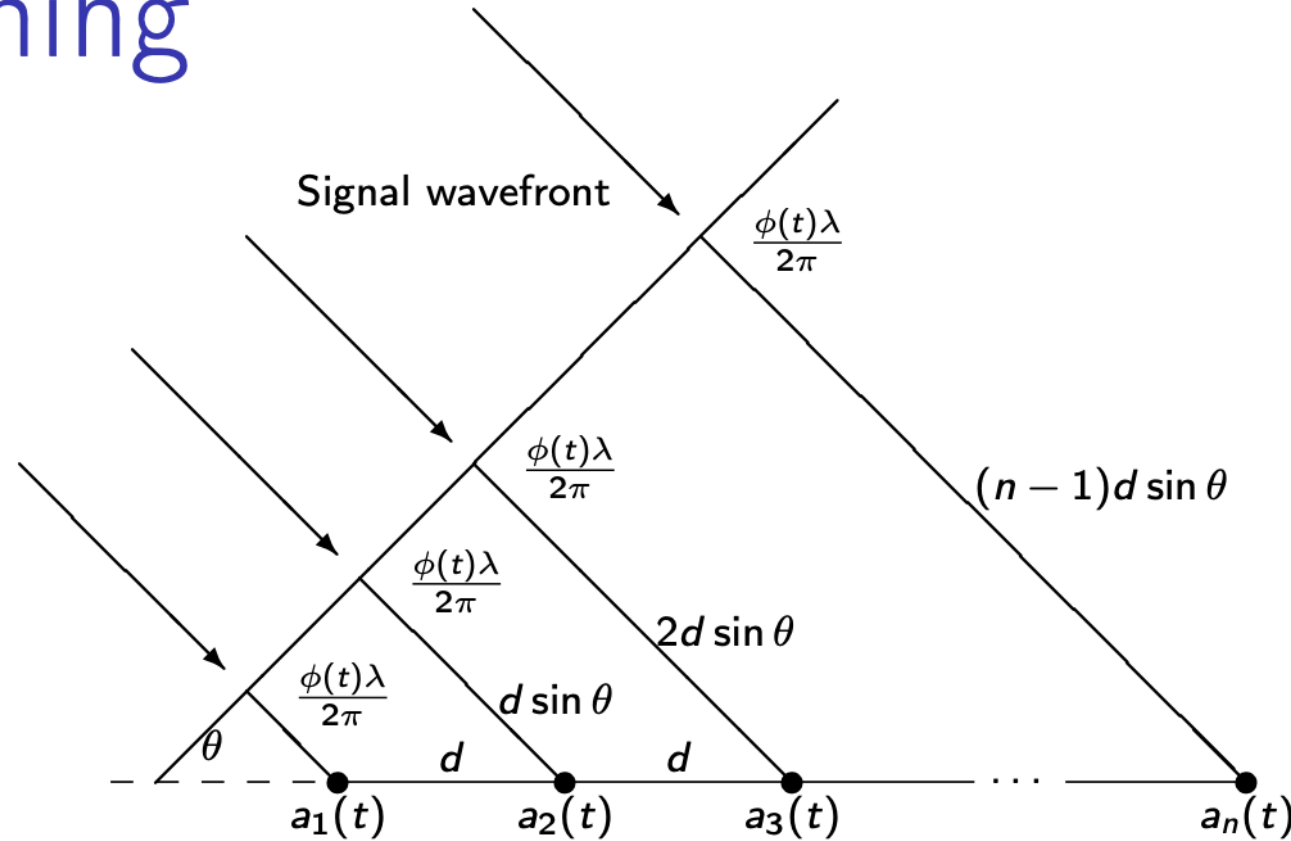
```
% the covariance matrix is defined as E{x.*conj(x)}  
if size(C,2) == 1  
    % MVD  
    temp = qrinsolve(x.',C);  
    w = G*temp/(C'*temp);  
else  
    % LCMV  
    if m >= n  
        [temp,F] = qrinsolve(x.',C);  
        w = temp*qrinsolve(F',G);  
    else  
        % when matrix is fat, F is no longer square and we cannot play the  
        % trick of thin matrix. Therefore, we have to form R2 and use LU.  
        temp = qrinsolve(x.',C);  
        R2 = C'*temp; % R2 = C'*R^(-1)*C  
        [L2, U2] = lu(R2);  
        temp2 = U2\((L2\G);  
        w = temp*temp2;  
    end  
end
```



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Beamforming



m samples, n antenna elements, $m \gg n$.

m -by- n data matrix A .

$a(t)$ is an n -by-1 column vector. $a(t)^H$ form the rows of A .

Unified notation

- A is an m -by- n data matrix
- $m \gg n$
- Beamformer problem: Solve $(A^H A)x = b$ where b is the steering vector.

Beamformer problem

1. Estimate correlation matrix: $m\mathcal{R}_{aa} = \sum_{t=1}^m a(t)a(t)^H = A^H A$
2. Solve $(A^H A)x = b$ where b is the steering vector.
3. Compute upper-triangular Cholesky factor of $(A^H A)$ and do forward and backward substitution.
4. QR vs. Cholesky
 - $R = \text{chol}(A^H A) \rightarrow A^H A = R^H R$
 - $[Q, R] = \text{qr}(A) \rightarrow QR = A \rightarrow A^H A = R^H Q^H QR = R^H R$

QR vs. Cholesky

Avoid computing $A^H A$ if you can. Never compute inverse.

Direct least-squares solution $Ax = b$ $(A^H A)x = b$	Normal equations least-squares $x = (A^H A)^{-1} A^H b$ $x = (A^H A)^{-1} b$
$R = \text{fixed.qlessQR}(A)$ 5.6648 2.3256 -0.8496 0 3.5967 -0.9131 0 0 2.4822	$R = \text{chol}(A' * A)$ 5.6648 2.3256 -0.8496 0 3.5967 -0.9131 0 0 2.4822

$$\text{fixed.qlessQR}(A) == [\sim, R] = \text{qr}(A, 0)$$

Minimum Variance Distortionless Response (MVDR) Beamformer

Matrix Solve Using Q-less QR Decomposition

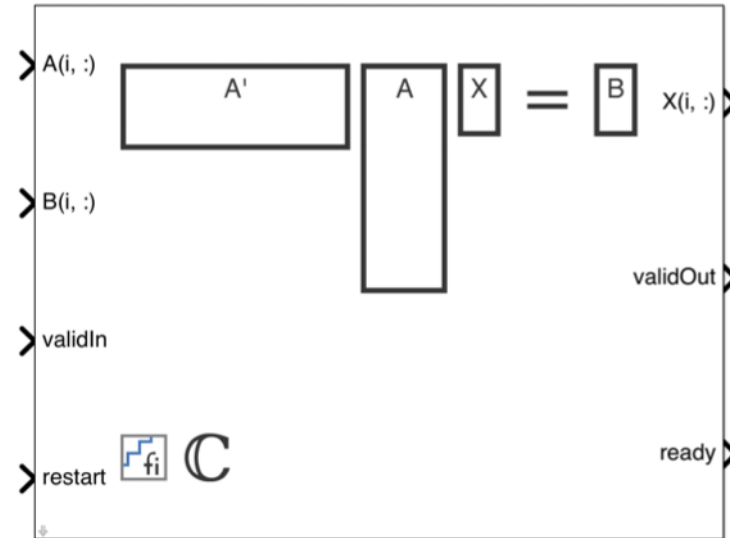
$$(A^H A)x = b$$

MVDR weight vector

$$w = \frac{X}{b^H X}$$

MVDR response

$$y = w^H a(t)$$



<i>Method</i>	<i>Input</i>	<i>Ready</i>	<i>Latency</i>	<i>Area</i>	<i>Release</i>
Burst	Row	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	$\mathcal{O}(n)$	R2020a
Partial-Systolic	Row	C	$\mathcal{O}(m)$	$\mathcal{O}(n^2)$	R2020b

MVDR with continuously streaming data

Matrix Solve Using Q-less QR Decomposition
with Forgetting Factor

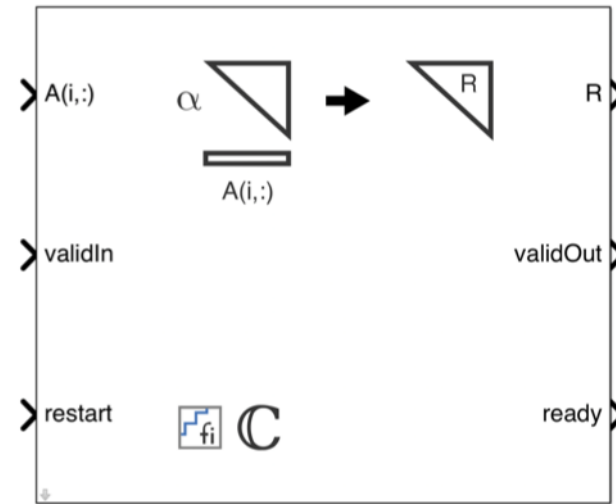
$$(A^H A)x = b$$

MVDR weight vector

$$w = \frac{x}{b^H x}$$

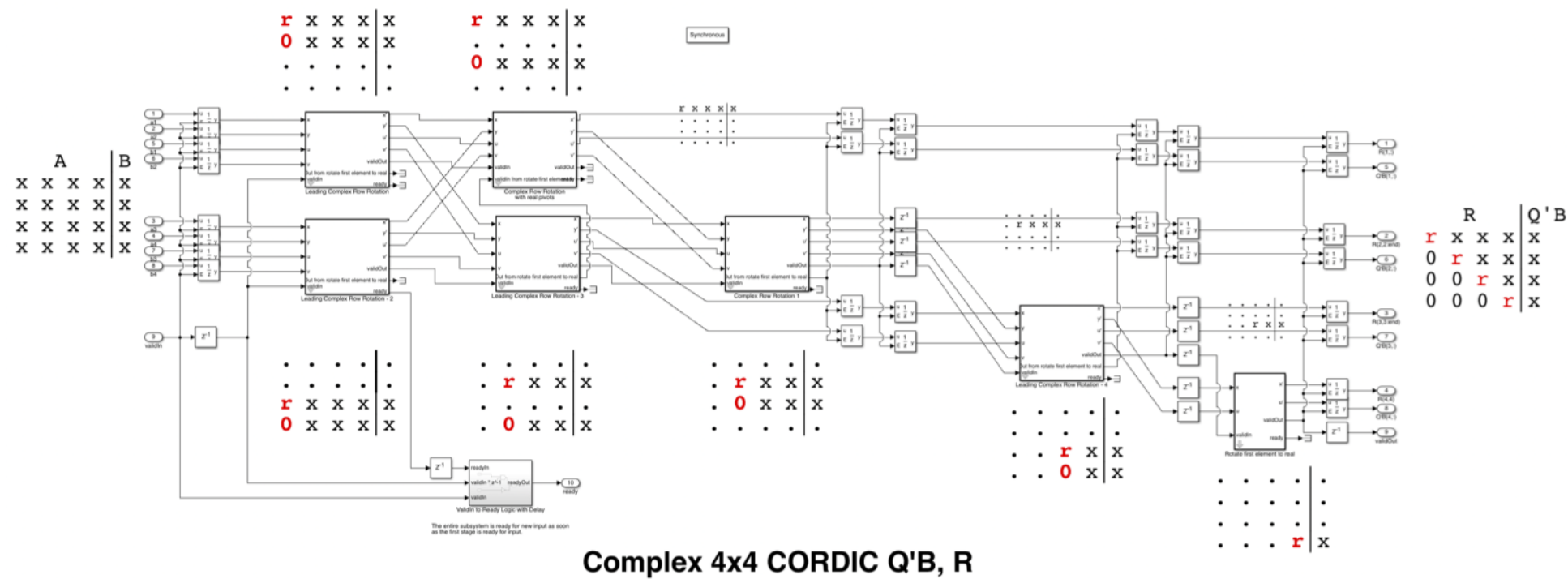
MVDR response

$$y = w^H a(t)$$



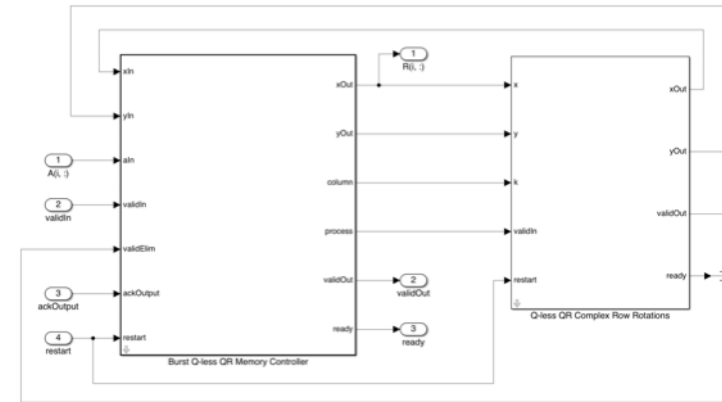
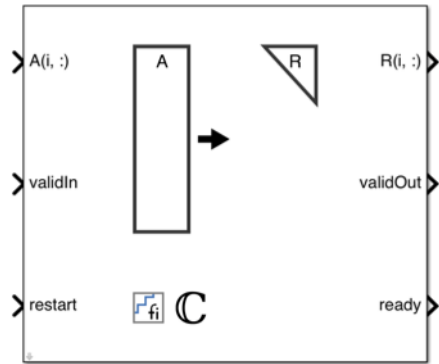
<i>Method</i>	<i>Input</i>	<i>Ready</i>	<i>Latency</i>	<i>Area</i>	<i>Release</i>
Partial-Systolic	Row	C	$\mathcal{O}(n)$	$\mathcal{O}(n^2)$	R2020b

Systolic: One cell for each zero ($\mathcal{O}(mn)$ cells). High area, Low latency.



	Method	Input	Ready	Latency	Area	Release
	Systolic	Matrix	C	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	R2019a Example
	Burst	Row	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	$\mathcal{O}(n)$	R2020a Library blocks
	Partial-Systolic	Row	C	$\mathcal{O}(m)$	$\mathcal{O}(n^2)$	R2020b Library blocks
	Partial-Systolic with Forgetting Factor	Row	C	$\mathcal{O}(n)$	$\mathcal{O}(n^2)$	R2020b Library blocks

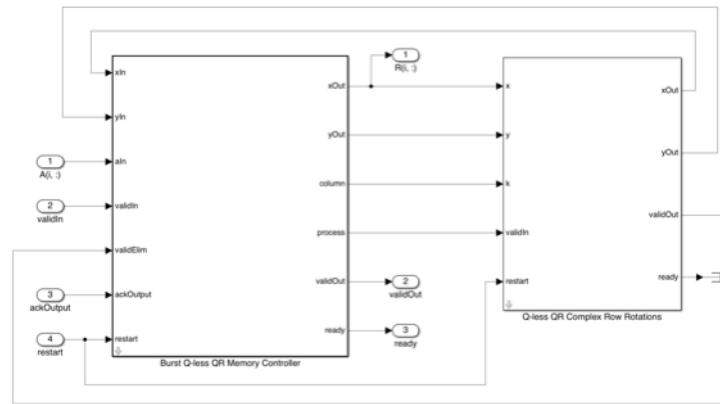
Burst: One cell for all zeros (1 cell). Low area, High latency.



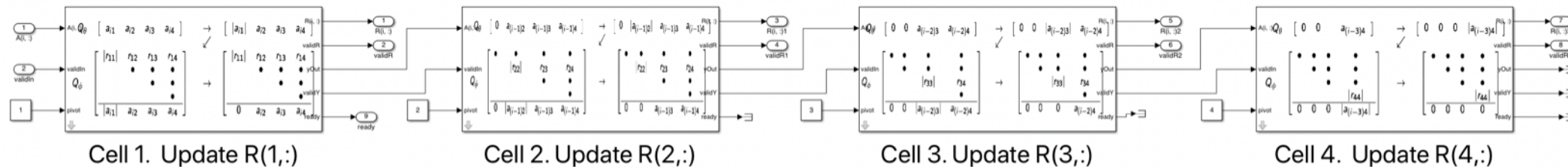
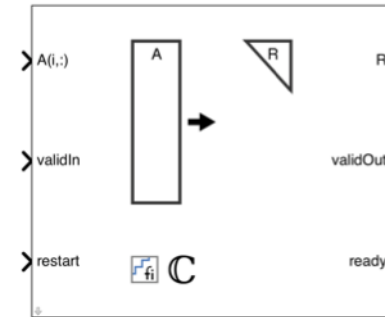
	Method	Input	Ready	Latency	Area	Release
	Systolic	Matrix	C	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	R2019a Example
	Burst	Row	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	$\mathcal{O}(n)$	R2020a Library blocks
	Partial-Systolic	Row	C	$\mathcal{O}(m)$	$\mathcal{O}(n^2)$	R2020b Library blocks
	Partial-Systolic with Forgetting Factor	Row	C	$\mathcal{O}(n)$	$\mathcal{O}(n^2)$	R2020b Library blocks

Partial-Systolic: (n cells). Medium area, Medium latency.

Cell internals



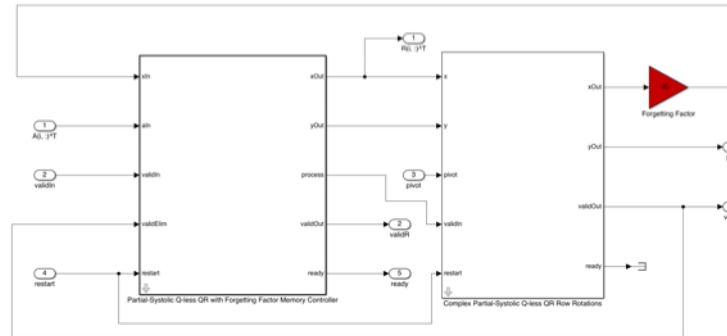
Block



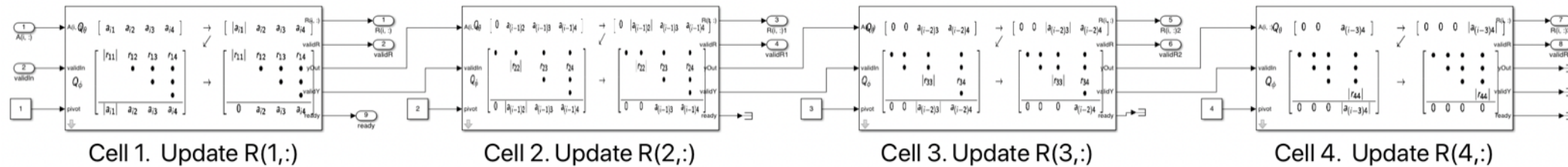
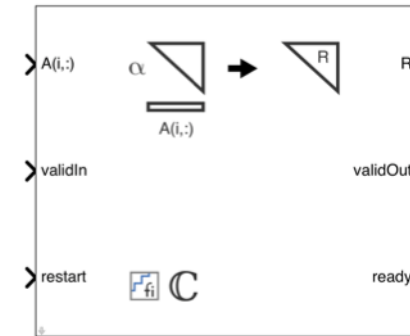
	Method	Input	Ready	Latency	Area	Release
	Systolic	Matrix	C	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	R2019a Example
	Burst	Row	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	$\mathcal{O}(n)$	R2020a Library blocks
	Partial-Systolic	Row	C	$\mathcal{O}(m)$	$\mathcal{O}(n^2)$	R2020b Library blocks
	Partial-Systolic with Forgetting Factor	Row	C	$\mathcal{O}(n)$	$\mathcal{O}(n^2)$	R2020b Library blocks

Partial-Systolic with Forgetting Factor (n cells): Continuously update

Cell internals with forgetting factor

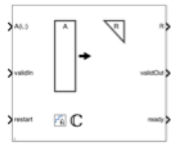


Block

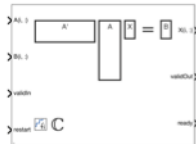


	<i>Method</i>	<i>Input</i>	<i>Ready</i>	<i>Latency</i>	<i>Area</i>	<i>Release</i>
	Systolic	Matrix	C	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	R2019a Example
	Burst	Row	$\mathcal{O}(n)$	$\mathcal{O}(mn^2)$	$\mathcal{O}(n)$	R2020a Library blocks
	Partial-Systolic	Row	C	$\mathcal{O}(m)$	$\mathcal{O}(n^2)$	R2020b Library blocks
	Partial-Systolic with Forgetting Factor	Row	C	$\mathcal{O}(n)$	$\mathcal{O}(n^2)$	R2020b Library blocks

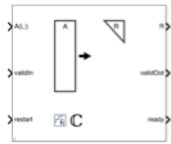
MATLAB functions



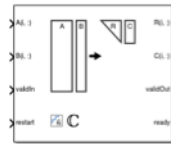
`fixed.qlessQR`



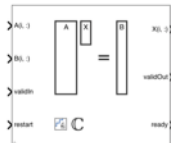
`fixed.qlessQRMatrixSolve`



`fixed.qlessQRUpdate`



`fixed.qrAB`



`fixed.qrMatrixSolve`

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 - MATLAB MVDR reference code
 - HDL Coder implementation
 - Hardware Prototyping – live demo

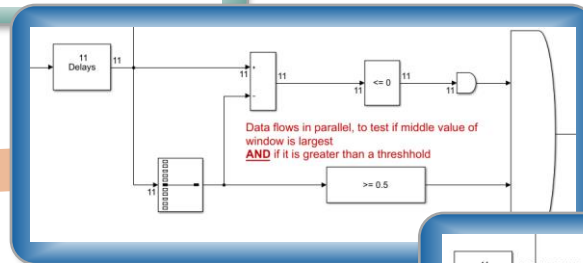
HDL Implementation Workflow

MATLAB
Reference

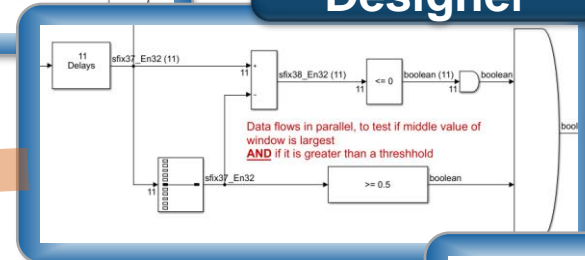
```
%% MATLAB reference detector
% this uses high level MATLAB functions
% computing a global maximum requires holding the entire signal at once
% this is impractical in a hardware implementation but serves as a golden
% reference

y=filter(CorrelationFilter,1,RxSignal); % correlate against the pulse
[peak, location]=max(abs(y).^2);
fprintf('Found Global Maximum at location %d Value %3.3f \n',location, peak)
```

Hardware
Architecture



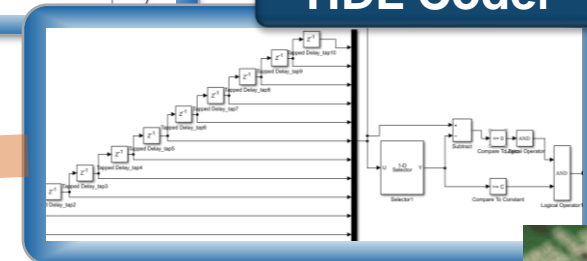
**Fixed Point
Designer**



Fixed-point
Implementation

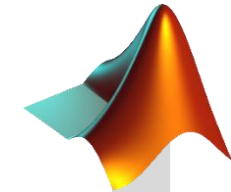
HDL Code Generation
and Optimization

HDL Coder

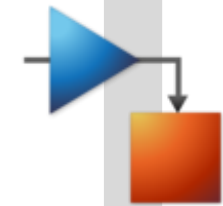


HDL Verification
and Targeting

Integrated Verification



MATLAB



Simulink



MATLAB MVDR reference code

```
function Y = mvdr_beamform(X, sv)
```

```
% form covariance matrix
```

```
EcX = X.'*conj(X);
```

```
% compute weight vector
```

```
wp = EcX\sv;
```

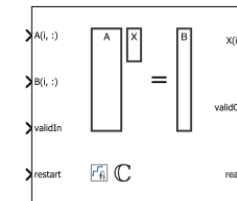
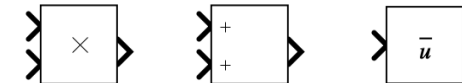
```
% normalize response
```

```
w = wp/(sv'*wp);
```

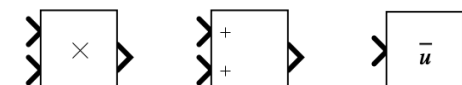
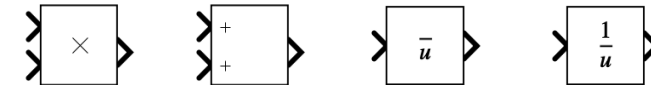
```
% form output beam
```

```
Y = X*w*conj(w);
```

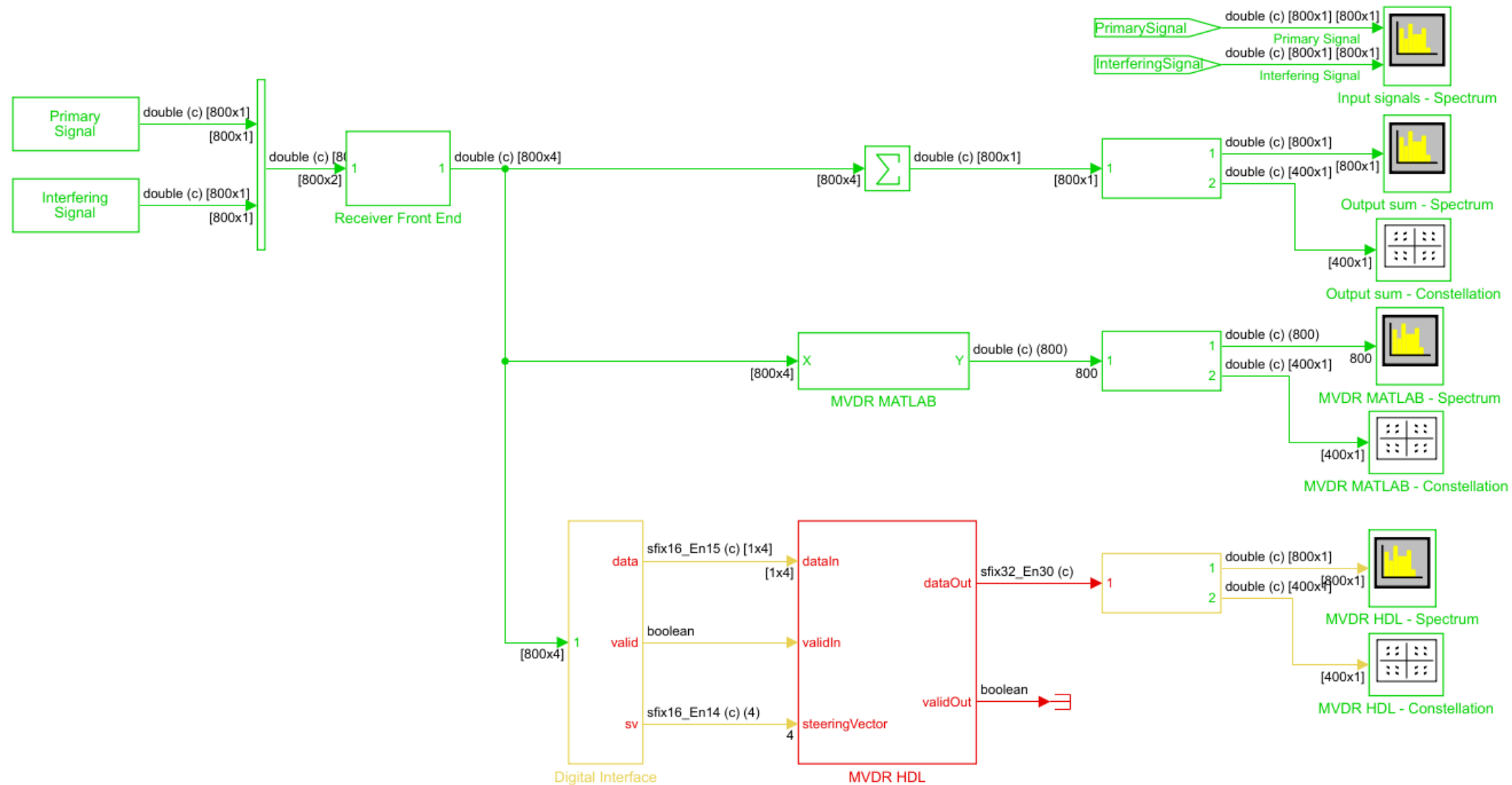
```
end
```



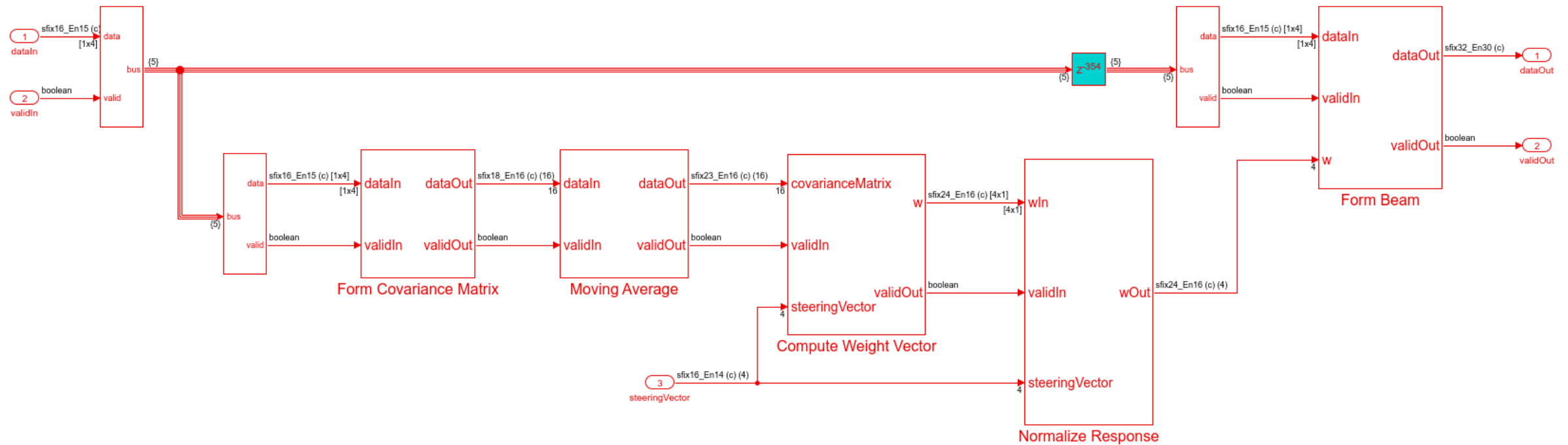
**100+ hours of
design time saved!**



HDL Implementation of MVDR Beamforming



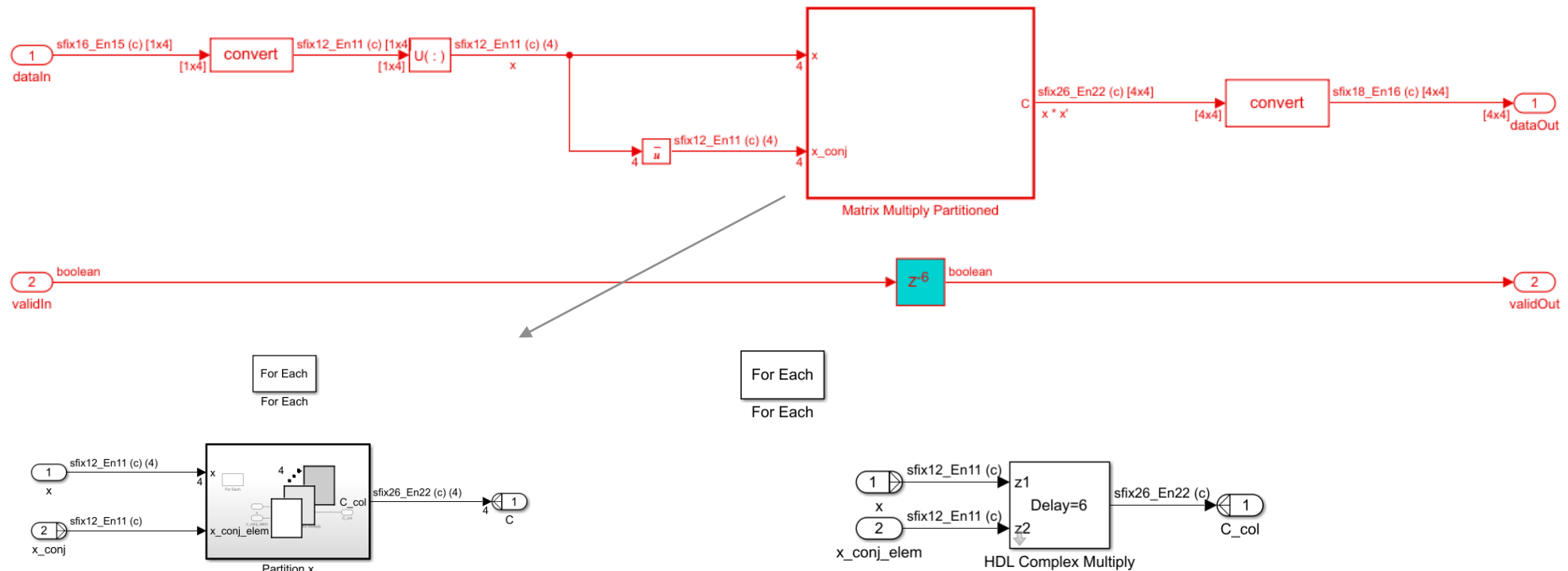
HDL Implementation of MVDR Beamforming



Form Covariance Matrix

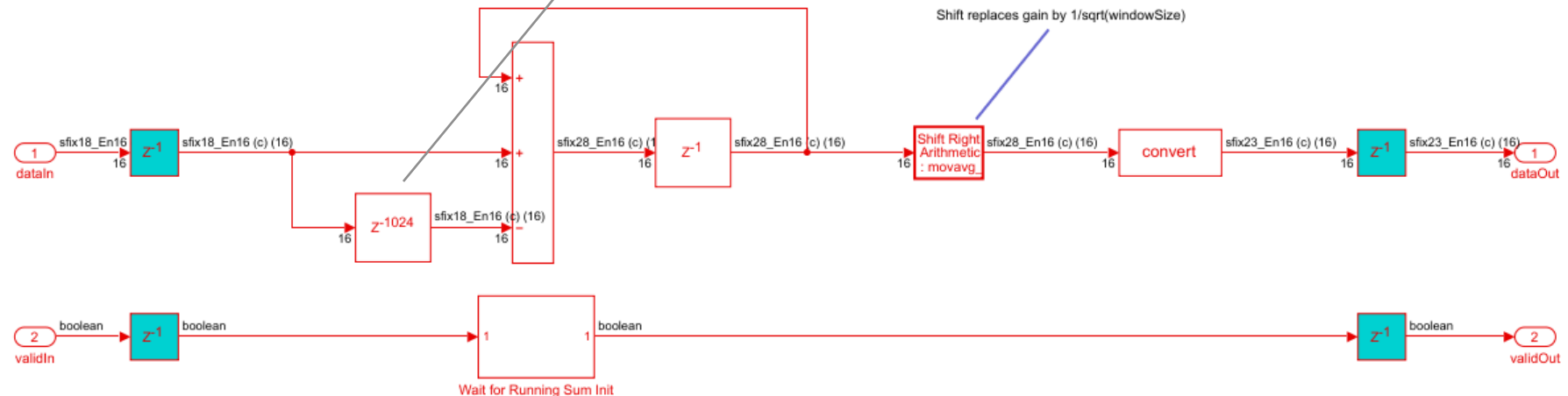
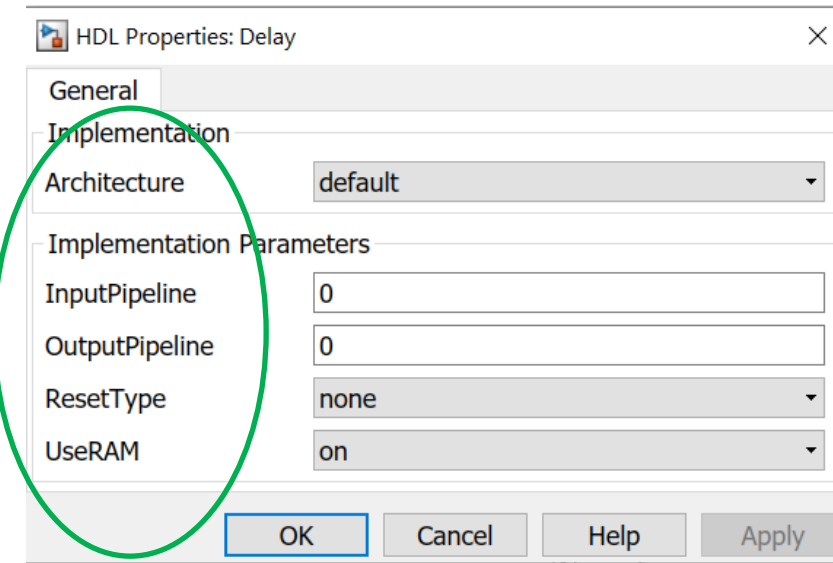
- For Each subsystem
 - Process elements independently
 - Concatenate results into outputs

```
% form covariance matrix
EcX = X.'*conj(X);
```



Moving Average

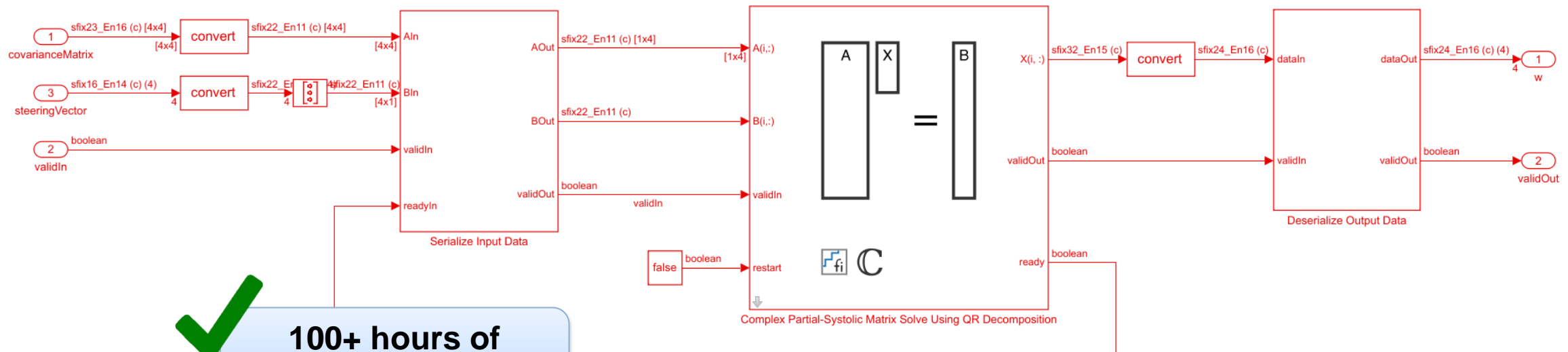
- Use HDL Implementation properties to map large delays to Block RAM



Compute Weight Vector

- Use Complex Matrix Solve block from Fixed-Point Matrix Linear Algebra Library

```
% compute weight vector
wp = Ecx\sv;
```

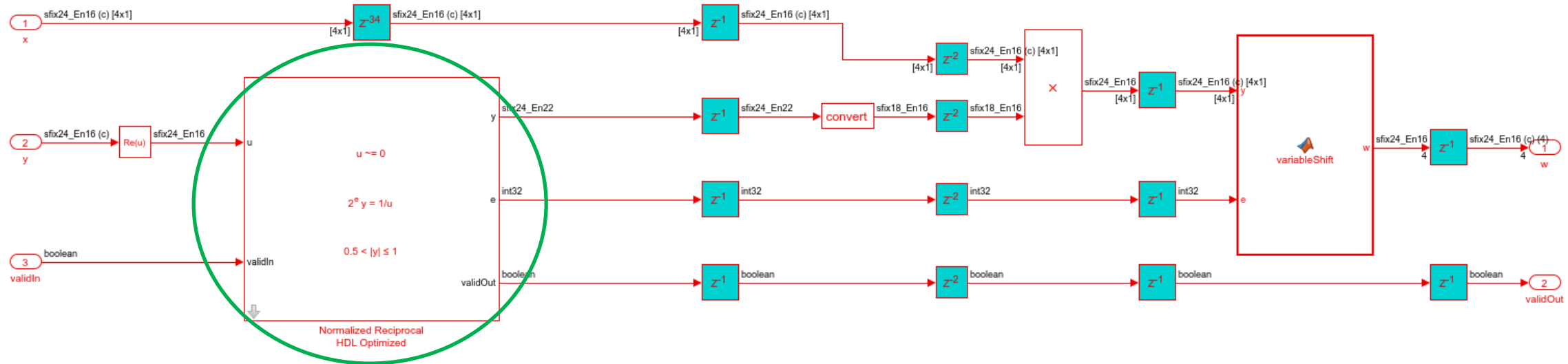


**100+ hours of
design time saved!**

Normalize Response

- Perform divide using reciprocal and multiply
- Fixed-point CORDIC reciprocal “just works”

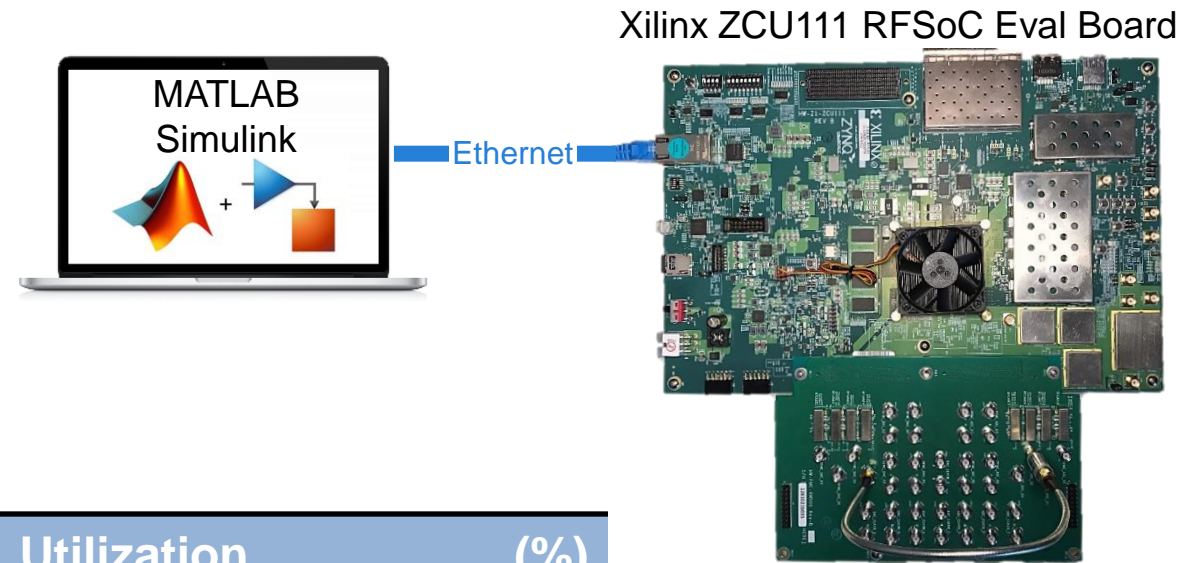
```
% normalize response
w = wp/(sv'*wp);
```



Implementation Results

- Device: xczu28dr (ZCU111)
- Maximum frequency: 452 MHz
- Resource utilization:

Resource	Utilization	(%)
LUT	47K	11.13
LUTRAM	989	0.5
FF	40K	4.7
BRAM	2	0.2
URAM	10	12.5
DSP	92	3.5



Resources to Get Started and Speed Adoption

- Getting started:
 - [MATLAB Onramp](#)
 - [Simulink Onramp](#)
 - [HDL pulse detector self-guided tutorial](#) and [videos](#)
- Proof-of-concept guided evaluations
 - **FREE** support via weekly WebEx meetings using custom sample designs
 - MathWorks coaches customers on “how to fish” through weekly WebEx sessions
- Training & consulting services
 - [HDL code generation](#), [FPGA signal processing](#) & [Zynq programming](#) training courses
 - Consulting service on deep technical coaching, custom design / hardware and more

INTRODUCING HDL CODER INTO WORKFLOW

Training period, ~2 months

- Learning Simulink
- Testing HDL Coder
- Finding limitations in synthesis
- Finding workarounds for limitations

Implementation, 1 month

- Design of custom components
- High level design of signal processing ch

Integration, 1 week

- Include generated code in FPGA framev
- Resolve timing issues

BENEFITS

- › A single model for simulation and code generation
 - No hand offs between systemization and implementation
 - Ready for FPGA as soon as simulation works
- › Short iterations for changes in design
- › Simulink block diagram resembles “manager level” power points
 - Drawback: makes it look a bit too easy
- › FPGA implementation done by person with limited VHDL competence

Tomas Andersson
Ericsson

Q&A