

# ESE532: System-on-a-Chip Architecture

Day 27: December 5, 2018  
Representation and Precision



## Today

- Fixed Point
- Errors from Limited Precision
- Precision Analysis / Interval Arithmetic
- Floating Point
  - If time permits

## Message

- We must always calculate with limited precision
- Precision costs area (and energy)
  - Can economize area (and energy) by judiciously using just the precision we need
    - Something can do when building customized accelerator
  - Precision-cost tradeoff → design-space axis
- Can perform analysis on precision

## Fixed Point

- Integer which interpret as a fraction
- Fixed-Point N.F
  - N bits
  - F bits below fraction (typically N>F)
  - Equivalently: meaning is Integer-value/2<sup>F</sup>
    - F=0 → Integer

$$A = \sum_{i=0}^{N-1} a_i 2^{i-F}$$

## Operator Sizes

Operator	LUTs	LUTs + DSPs
Double FP Add	712	681+3 DSPs
Single FP Add	370	219+2 DSPs
Fixed-Point Add (32)	16	
Fixed-Point Add (n)	n/2	
Double FP Multiply	2229	223+10 DSPs
Single FP Multiply	511	461+3 DSPs
Fixed Multiply (32x32)	1099	
Fixed Multiply (16x16)	283	1 DSP
Fixed Multiply (18x25)		1 DSP
Fixed Multiply (n)	~ n <sup>2</sup>	

FP (Floating Point) sizes from:  
[https://www.xilinx.com/support/documentation/ip\\_documentation/ru/floating-point.html](https://www.xilinx.com/support/documentation/ip_documentation/ru/floating-point.html)

## Observe

- Floating-Point operators are large compared to Fixed-Point
  - For similar precision
    - 712 vs. 32 for addition
- Double-precision Floating point operators are large
  - 2229 Multiply, 712 Add
    - Can quickly fill 50,000 LUT programmable logic

## Fixed-Point Economy

- Can fit more logic (more parallelism) using modest fixed-point
  - At 16b: Multiply 283, Add 16
  - Vs. Double: 2229, 712
- But
  - How much precision do we need?
  - How do we determine?

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## Perfect Representation

- Start with Fixed-Point 16.8
- What do we need to
  - represent the result of an addition? (1a)
  - represent the result of a multiplication? (1b)

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## Sequence

- Across a sequence of operations
  - A, B, C start Fixed-Point 16.8
- $Y=(A+B)*C$
- Perfect representation for partial results up to Y?

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## Looping: Bound loop

```
res=0;
for(i=0;i<3;i++)
  res=res*x+a[i]
```

- Assume  $a[i]$ ,  $x$  start Fixed-Point 16.8
- Final precision needed for  $res$ ?

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## Looping: Unbound

```
res=0;
for(i=0;i<len;i++)
  res=res*x+a[i]
```

- Assume  $a[i]$ ,  $x$  start Fixed-Point 16.8,  $len$  starts Integer 16
- Final precision needed for  $res$ ?

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## Perfect Representation

- Start with Fixed-Point 16.8
- What do we need to
  - represent the result of a division? (1c)
    - E.g. 00000001.00000000/00000011.00000000

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## Conclude

- Cannot generally keep perfect precision
- Will typically need to decide how much precision we need and where

## Errors from Limited Precision

Accept errors necessary.  
How big are they?  
How design to manage them?

## What error introduce?

- **Absolute Error** – what level of error do we have in approximated value or a result
- Might be all we care about
  - Get answer to 1mV accuracy
  - ...or 1 pixel accuracy

- 4.13742 – assume full
- 4.1374 – 0.00002 –  $10^{-5}$
- 4.137 – 0.00042 –  $10^{-4}$
- 4.14 – 0.00256 –  $10^{-3}$
- 4.1 – 0.03742 –  $10^{-2}$

## What error introduce?

- **Relative Error** – error as percentage of intended result
- May be more relevant, particularly if trying to identify small values

- 4.13742 – 0.0003
  - Ideal: 4.13712
  - 2 decimal frac: 4.14-0.00
  - 4.14
  - $(4.14-4.13712)/4.13712$
  - 0.07% error
- 4.13742-4.13628
  - Ideal: 0.00114
  - 2 decimal fac: 4.14-4.14
  - 0.00
  - $(0.00114-0)/0.00114$
  - 100% error

## Preclass 2

- Complete Table

Reduced Precision Calculation	Y	Y <sub>1</sub>	Error (2 significant figures)	
			Absolute $ Y - Y_1 $	Relative $ (Y - Y_1)/Y $
$Y_1 = A_1 + B_1$				
$Y_1 = A_1 \times B_1$				
$Y_1 = A_1/C_1$				
$Y_1 = A_1/D_1$				

## Observe

- Add/Multiply relatively well behaved
- Must be very careful when
  - Division involved
  - Divisors can be small
    - Get approximated near zero

## Precision Allocation

- Full precision can be too expensive
  - Non-sensical
- Limited precision introduces errors
  - May be smaller than we care about
- Determine minimal precision needed
  - ...or where to spend precision...

## Empirical Analysis

- Make guess at precisions
- Set precisions in calculation
- Simulate on data
- Evaluate results (absolute, relative error) compared to gold standard
  - Unlimited precision...or, at least, higher precision
    - Often standard is double-precision float
    - ...but, as we'll, even that's a compromise
- Update precision guess and repeat

## Empirical Analysis

- Make guess at precisions
  - Set precisions in calculation
  - Simulate on data
  - Evaluate results compared to gold standard
- Demands Care
- Test coverage
    - Adequate set of test data to trigger worst-case errors?
  - Requires some understanding of calculation
  - Shouldn't be entirely black box

## Vivado HLS Support

- Has libraries to support
  - Arbitrary precision integers
  - Arbitrary precision fixed point
- For
  - Simulation
  - Synthesis
- UG902 – Vivado HLS User Guide
  - Chapter 2: Arbitrary Precision Data Type Library

## Precision Analysis

- Can analyze worst-case error impacts from limited precision
- Give results not sensitive to test set
- Give guidance on where to allocate precision
- ...can be automated

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## Limit Precision Inputs

- Typically start with limited precision
  - A/D only sample to 12b
    - Real-world had more precise value, but didn't capture
  - Quantized data stored in representation
    - Sound samples, DCT frequency coefficients
- We start with error
  - What does that mean about values we calculate?

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## Interval Analysis

- Treat every value as an interval arrange
- Model effects of operations on range of results
- $A=(A.H, A.L)$  e.g. read 37 (37.49,37.51)
- $A+B=(A.H+B.H, A.L+B.L)$
- Assuming Positive A, B and interval not cross 0, what is range for:

- $A*B$
- $A/B$

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## Interval Analysis

- Treat every value as an interval arrange
- Model effects of operations on range of results
- $A=(A.H, A.L)$
- $A+B=(A.H+B.H, A.L+B.L)$
- Positive A, B and B interval not cross 0
  - $A*B=(A.H*B.H, A.L*B.L)$
  - $A/B=(A.H/B.L, A.L/B.H)$

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## Interval Analysis

With ranges that may cross zero...

- $A*B=(\max(A.H*B.H, A.H*B.L, A.L*B.H, A.L*B.L), \min(A.H*B.H, A.H*B.L, A.L*B.H, A.L*B.L))$
- $A/B$  ... more complicated
  - If B.H positive, B.L negative, can become infinity

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## Limited Precision

- $A=(A.H, A.L)=(A+\epsilon, A-\epsilon)=A\pm\epsilon$
- E.g. if rounded to Fixed Point 16.8
  - $\epsilon$  may be  $2^{-9}$

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## Preclass 2

- Complete Table

Calculation	Result Range (report like $A$ )
$Y = 1 + A$	$A_g + 1 \pm \epsilon$
$Y = A + B$	
$Y = 2A$	
$Y = A \times B$	

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## Multiplication

- $(A \pm \epsilon)(B \pm \epsilon)$
- $A * B \pm ((A+B) \epsilon + \epsilon^2)$
- A and B can be MAXVAL, MINVAL
  - Assume symmetric (MAXVAL=-MINVAL)
- $A * B \pm (2 * \text{MAXVAL} * \epsilon + \epsilon^2)$ 
  - Probably reasonable to drop  $\epsilon^2$
- $A * B \pm 2 * \text{MAXVAL} * \epsilon$

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## Multiply Range Example

- Recall: Fixed-Point 16.8 multiply
  - Full precision result: 32.16 (preclass 1b)
- What is error interval for result of 16.8 multiply?

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## Range Example

- Error Interval for 16.8 fixed-point multiply
- For 16.8,  $\epsilon_0 = 1/512$ , maxval=256
- From preclass 3 symbolic answer
  - $A * B \pm 2 * \text{MAXVAL} * \epsilon$
  - $A * B \pm 2 * 256 * (1/512)$
  - Or  $A * B \pm 1$

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## Observe

- Full precision may keep bits well below the error in the calculation
  - E.g. 32.16 result, keeping 16b below  $2^0$  term
  - Entire fraction is below the error in the calculation
    - $A * B \pm 1$

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## Rounding

- Rounding introduces an error
- Round Nearest A to Fixed-Point N.8
  - $\epsilon = 2^{-9}$
- As does truncation, floor, ceil...
  - But asymmetric interval

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## Compute and Round Error

- Error range if round 32.16 result to 18.2?
  - (from 16.8 multiply)
  - Hint:
    - How much from calculation?
    - How much additional from rounding?

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## Compute and Round Error

- Rounding 32.16 to 18.2 introduces a quantization error of  $2^{-3}$
- Our 32.16 multiply result was  $\pm 1$ 
  - Add an addition  $\pm 1/8$
  - Total error:  $\pm 9/8$

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## Result

- Dropping (rounding) bits may not increase error range (much)

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## Symbolic

- If work through computation symbolically, can generate equation for error
- Each rounding (precision drop) adds some  $\epsilon_i$  precision

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## Approach

- At each step compute interval
- Keep track of
  - maxval
  - $\epsilon$

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## Symbolic Example

- Start A, B, C with  $\text{maxval}_0, \epsilon_0$
- $Y=(A+B)*C$ ;
- $A+B$   $\text{maxval}_1=2\text{maxval}_0, \epsilon_2=2\epsilon_0+\epsilon_1$ 
  - $\epsilon_1$  for round after this operation
- $(A+B)*C$ 
  - $\text{maxval}_2=\text{maxval}_1*\text{maxval}_0$
  - $\epsilon_4=\epsilon_2*\text{maxval}_0+\epsilon_0*\text{maxval}_1+\epsilon_3$
  - $\epsilon_3$  for round at this operation

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## Result Precision

- After series of operations, may have expression like:
  - $Y \pm (\epsilon_3 + (\epsilon_1 + 4 * \epsilon_0) * \text{maxval}_0)$
- If looking for result with precision  $\pm \epsilon_{\text{res}}$ 
  - $\epsilon_{\text{res}} \geq (\epsilon_3 + (\epsilon_1 + 4 * \epsilon_0) * \text{maxval}_0)$

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## Result Precision

- Fixed Precision  $12.0 = \text{Round}(\text{val})$ 
  - E.g. A/D output
  - Only need to know val to  $\epsilon=1/2$
- Fixed Precision  $12.0 = \text{Round}(\text{val}/4)$ 
  - E.g. Quantized value stored in file
  - Only need to know val to  $\epsilon=2$
  - Start 13.8,  $\text{maxval}_0=32$
  - $\epsilon_{\text{res}} \geq (\epsilon_3 + (\epsilon_1 + 4 * \epsilon_0) * \text{maxval}_0) = (\epsilon_3 + (\epsilon_1 + 4 * \epsilon_0) * 32)$ 
    - What epsilons might solve?
    - Hint: try budget half unit for each  $\epsilon_3, \epsilon_1$  term

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## Optimize Precision Allocation

- $\epsilon_{\text{res}} \geq (\epsilon_3 + \epsilon_1 + 4 * \epsilon_0 * 32)$ 
  - Maybe:  $\epsilon_0=1/256, \epsilon_1=1/2, \epsilon_3=1/64$
  - $1/256$  – 7 bit fraction,  $1/64$  – 5 bit,  $1/2$  – no fraction
- More generally
  - Combine with area model and look at expense of providing each  $\epsilon_i$
  - Round to 12.7, Fixed 12.7 add, round to 11.5, Fixed 11.5 multiply, round to 12.0
    - $12/2$  add,  $12/2$  round,  $11 * 11$  for multiply ~ 133 LUTs
- Try pick  $\epsilon_i$  to meet  $\epsilon_{\text{res}}$  while minimizing area

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## Tools

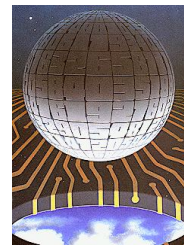
- Tools can automate interval calculations to verify precision
  - E.g. Gappa++
  - <https://bitbucket.org/mlinderm/gappa>

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## Floating Point

(time permit)



Robert Tinney  
(Byte circa 1980)

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## Floating-Point vs. Fixed-Point

- Floating-Point (esp. double-precision) is a **big hammer** solution
  - Trades hardware/energy for programmer attention to needs
  - Standards have been well thought out so works over wide range
    - One size fits all (...almost)
  - Most cases it is more than needed
- Cost/energy sensitive designs will ask what's necessary and tune accordingly

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## Floating Point

- Leading 0s aren't that useful
- Can represent more compactly by counting them
  - Only need log bits to count the zeros
- Represent value as  $v=1.m * 2^{(\text{exp}-\text{offset})}$ 
  - Mantissa (m)
  - Exponent (exp)
- Like Scientific Notation

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## Floating Point

- Floating Point means
  - Move the datapath to the interesting/significant part of computation
  - Don't represent leading zeros
  - Don't represent less significant bits
    - Even if they are well above 1
      - In the integer portion

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## Standard Floating Point

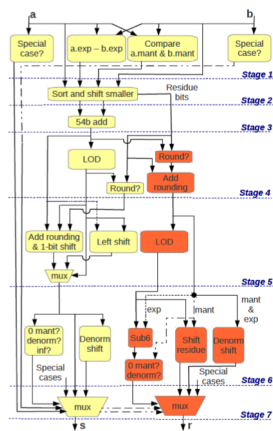
- Double Precision (64b)
  - 1 bit sign
  - 11 bit exponent
    - Offset 1023 represents 1023 to -1022
  - 53 bit mantissa (52b + implicit 1)
- Single Precision (32b)
  - 1 bit sign
  - 8 bit exponent (offset 127)
  - 24 bit mantissa (23b + implicit 1)

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## Expensive FP Add

- Recall
  - 712 LUTs double
  - 370 single
- Double:
  - 54b add one stage of 7 in pipeline
  - Done in 27 6-LUTs



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## Floating Point Multiply

Double 2229, Single 461

- Don't need to sort, pre-shift
- $m=A.m*B.m$  (53x53 multiply dominates)
- $e=A.e+B.e$
- Still have shifting, rounding at end

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## Dynamic Range

- Floating Point has very wide dynamic range
- Can deal with significant piece being anywhere in 1023 to -1022
- For fixed-point to cover same range
  - Fixed Point 2046.1022
  - Add 1024 LUTs
  - Multiply ~ 4M LUTs
- When need dynamic range, FP economical

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## Customization

- Can customize Floating Point on FPGA or custom silicon
  - Mantissa bits
  - Exponent bits
- Fewer bits when need less precision or range to save area
- More bits if need greater precision or range

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## Floating Point

Not free of precision problems

- $((1+2^{100})-2^{99})-2^{99}=0$
- $1+(2^{100}+(-2^{99}-2^{99}))=1$

## Floating-Point Analysis

- Can do similar analysis on floating point
  - ...and there are tools to help
  - Including Gappa++

## Big Ideas

- We must always calculate with limited precision
- Precision costs area (and energy)
  - Can economize area (and energy) by judiciously using just the precision we need
  - Precision-cost tradeoff
    - Design-space axes
- Can perform analysis on precision

## Admin

- Project Due Friday
  - Report individual
  - Elf, bitstream, decoder – one per group
  - Code – everyone turn in, but same across group
- Return boards Monday in Class
- Exam following Friday (12/14)
  - Towne 303 (here), 9am