Floating Point Arithmetic

1 Introduction

Fixed point numbers suffer from limited range and accuracy. For a given word length both fixed point and floating point representations give equal distinct numbers. The difference is that in fixed point representation the spacing between the numbers is equal, so smaller numbers when truncated or rounded give a much larger error than the larger numbers. However floating point representation gives different spacing between numbers. We get denser distances between numbers when the number is small and sparser distance for larger numbers. So the absolute representation error increases with larger numbers.

Floating point numbers are used to obtain a dynamic range for representable real numbers without having to scale the operands. Floating point numbers are approximations of real numbers and it is not possible to represent an infinite continum of real data into precisely equivalent floating point value.

Number system is completely specified by specifying a suitable base β, significand (or mantissa) *M*, and exponent *E*. A floating point number *F* has the value

$F=M\ \beta^{E}$

β is the base of exponent and it is common to all floating point numbers in a system. Commonly the significand is a signed - magnitude fraction. The floating point format in such a case consists of a sign bit *S*, *e* bits of an exponent *E*, and *m* bits of an unsigned fraction *M*, as shown below

S	Exponent *E*	Unsigned Significand *M*

The value of such a floating point number is given by:

$F = (-1)^{S} M \beta^{E}$

The most common representation of exponent is as a biased exponent, according to which

$E = E^{true} + bias$

bias is a constant and E^{true} is the true value of exponent. The range of E^{true} using the *e* bits of the exponent field is

$$-2^{e-1} \le E^{true} \le 2^{e-1} - 1$$

The bias is normally selected as the magnitude of the most negative exponent; i.e. 2^{e-1}, so that

$$0 \le E \le 2^{e} - 1$$

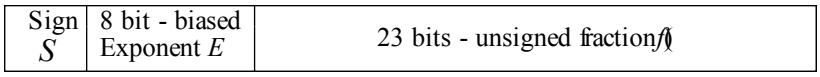
The advantage of using biased exponent is that when comparing two exponents, which is needed in the floating point addition, for example the sign bits of exponents can be ignored and they can be treated as unsigned numbers

The way floating point operations are executed depends on the data format of the operands. IEEE standards specify a set of floating point formats, viz., single precision, single extended, double precision, double extended. Table 1 presents the parameters of the single and double precision data formats of IEEE 754 standard.

Fig.2 1 shows the IEEE single and double precision data formats. The base is selected as 2. Significands are normalized in such a way that leading 1 is guaranteed in precision (p) data field. It is not the part of unsigned fraction so the significand is in the form *1.f*. This increases the width of precision, by one bit, without affecting the total width of the format.

Table 1: Format parameters of IEEE 754 Floating Point Standard

Parameter	Format	
	Single Precision	Double Precision
Format width in bits	32	64
Precision (p) = fraction + hidden bit	23 + 1	52 + 1
Exponent width in bits	8	11
Maximum value of exponent	+ 127	+ 1023
Minimum value of exponent	-126	-1022



(a) IEEE single precision data format

(b) IEEE double precision data format

Fig 2.1 - Single and double precision data formats of IEEE 754 standard

The value of the floating point number represented in single precision format is

$$F = (-1)^S 1.f 2^{E-127}$$

where 127 is the value of *bias* in single precision format ($2^{n-1}-1$) and exponent E ranges between 1 and 254, and E = 0 and E = 255 are reserved for special values.

The value of the floating point number represented in double precision data format is

$$F = (-1)^S 1.f 2^{E-1023}$$

Where1023 is the value of *bias* in double precision data format. Exponent E is in the range.

$$1 \leq E \leq 2046$$

The extreme values of E (i.e. $E = 0$ and $E = 2047$) are reserved for special values.

The extended formats have a higher precision and a higher range compared to single and double precision formats and they are used for intermediate results [2].

2. Choice of Floating Point Representation

The way floating point operations are executed depends on the specific format used for representing the operands. The choice of a floating point format for the hardware implementation of floating point units is governed by factors like the dynamic range requirements, maximum affordable computational errors, power consumption etc. The exponent bit width decides the dynamic range of floating point numbers while the significand bit

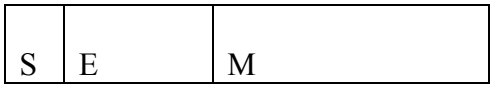
width decides the resolution. The dynamic range offered by floating point units is much higher than that offered by fixed point units of equivalent bit width. Larger dynamic range is of significant interest in many computing applications like in multiply - accumulate operation of DSPs. But larger range is not needed in all the applications. The actual bit-width required in many applications need not match with the ones provided by IEEE standards. For example, considering the design of an embedded system, the choice of IEEE data formats need not give optimal results. In some cases, even IEEE specified rounding scheme may not guarantee acceptable accuracy. That means, depending on the specifications of a certain application, dedicated system solutions can work with non IEEE data formats as well as rounding schemes such that the real life input/output signals satisfy the data processing goals required by the target application. Although custom specification of floating point format do find some applications, in the recent years more and more manufacturers are following IEEE standards for the design of their hardware. IEEE compliance guarantees portability of software between different platforms. Applications that do not need such portability need not stick to IEEE standards.

Examples_1:

For an 8 bit word, determine the range of values that it represents in floating point and the accuracy of presentation for the following scenarios: (Assume a hidden 1 representation and extreme values are not reserved).

a) If 3bits are assigned to the exponents

b) If 4 bits are assigned to the exponents



Answer:

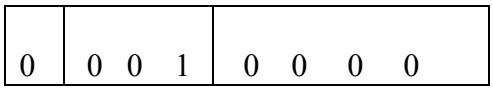
a) S=0, E=3bits, M=4bits,

Then the bias is $2^{n-1} - 1 = 2^{3-1} - 1 = 3$

Maximum range,

$$(-1)^0\ 1.1111\ 2^{7-3} = 1.1111\ 2^4 = 11111 = 31_{10}$$

Minimum range, assuming exponent 000 is reserved for zero

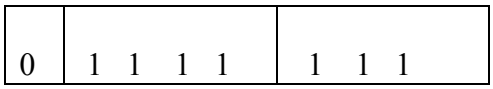


$$(-1)^0\ 1.0000\ 2^{1-3} = 1.0000\ 2^{-2} = 0.01 = 0.25_{10}$$

b) S=0, E=4bits, M=3bits ,

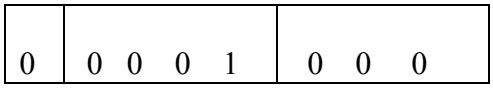
Then the bias is $2^{n-1} - 1 = 2^{4-1} - 1 = 7$

Maximum range,



$$(-1)^0\ 1.111\ 2^{15-7} = 1.111\ 2^8 = 111100000 = 480_{10}$$

Minimum range, assuming exponent 000 is reserved for zero



$$(-1)^0\ 1.0000\ 2^{1-7} = 1.0000\ 2^{-6} = 0.000001 = 0.015625_{10}$$

You must know that the total No. of numbers that can be represented is the same. The difference between example (a) and example (b) is that the resolution of the numbers that can be represented is different.

Exercise:

For a) above determine the decimal number corresponding to when M contains 0001 and 0010.

For b) above determine the decimal number corresponding to when M contains the number 001 and 010.

Discuss the resolution of the numbers represented in (a) & (b).

Example_2

Represent 21.75_{10} in Floating point. Use the IEEE 754 standard.

Answer:

21.75 in binary is 10101.11 or $1.010111\ 2^4$

S=0

Bias is 2^7-1=127 E= 127 + 4 = 131

1bit	8 bits	23 bits
0	1 0 0 0 0 0 11	0 1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Example _3

Represent -0.4375_{10} in floating point, using IEEE standard 754

Answer:
Binary equivalent of –0.4375 = -.0111 or $-1.11\ 2^{-2}$

S= 1
Exponent is –2 + 127 =125 or 01111101

1bit	8 bits	23 bits
1	01 1 1 1 1 01	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

3 IEEE Rounding

All real numbers can not be represented precisely by floating point representation. There is no way to guarantee absolute accuracy in floating point computations. Floating point numbers are approximations of real numbers. Also the accuracy of results obtained in a

floating point arithmetic unit is limited even if the intermediate results calculated in the arithmetic unit are accurate. The number of computed digits may exceed the total number of digits allowed by the format and extra digits have to be disposed before the final results are stored in user-accessible register or memory. When a floating point number has to be stored or output on the bus, then the width of the memory and the bus dictates that certain numbers greater than the width of the significand to be removed. Rounding is the process of removing the extra bits with the digital system resulting from internal computation (higher precision) to the exact bus width. IEEE 754 standard prescribes some rounding schemes to ensure acceptable accuracy of floating point computations. The standard requires that numerical operations on floating point operands produce rounded results. That is, exact results should be computed and then rounded to the nearest floating point number using the 'round to nearest - even' approach. But in practice, with limited precision hardware resources, it is impossible to compute exact results. So two guard bits (*G* and *R*) and a third bit, sticky (*S*), are introduced to ensure the computation results within an acceptable accuracy using minimum overhead.

The default rounding mode specified by the IEEE 754 standard is round to nearest - even. In this mode, the results are rounded to the nearest values and in case of a tie, an even value is chosen. Table 2.2, shows the operation of round to nearest - even, for different instances of significand bit patterns. In this table, X represents all higher order bits of the normalized significand beyond the LSBs that take part in rounding while the period is

Table 2.2: Round to nearest - even rounding

Significand	Rounded Result	Error	Significand	Rounded Result	Error
X0.00	X0.	0	X1.00	X1.	0
X0.01	X0.	- 1/4	X1.01	X1.	- 1/4
X0.10	X0.	- 1/2	X1.10	X1. + 1	+ 1/2
X0.11	X1.	+ 1/4	X1.11	X1. + 1	+ 1/4

separating *p* MSBs of the normalized significand from round (*R*) and sticky (*S*) bits. It can be seen from the table that the average bias (which is the average of the sum of errors for all cases) for the round to nearest scheme is zero. Fig 2.2 illustrates the relative positions of the decision making bits. Rounding to the nearest value necessitate a conditional addition of 1/2 *ulp* (units in the last place). The decision for such addition can be reached through the evaluation of the LSB (M_0) of the most significand *p* bits of the normalized significand, the round (*R*) bit and the sticky (*S*) bit. Rounding is done only if condition $R.(M_0 + S)$ is true (Boolean).

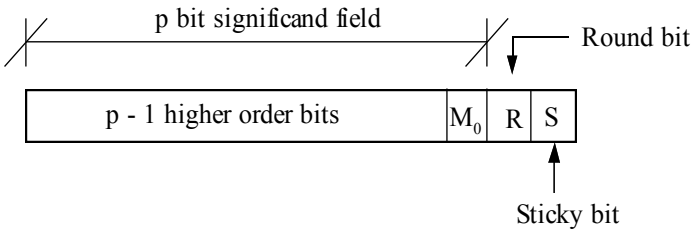


Figure 2.2 - Normalized Significand before rounding

Example:

Round the following data structure, according to the nearest even

			M_0	R	S
0	01 1 1 1 1 01	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1	1	1	

Answer:

Since R(M0 + S) holds, then, the rounding will produce the following results:

0	01 1 1 1 1 01	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0

4 Floating Point Multiplication

The algorithm of IEEE compatible floating point multipliers is listed in Table 2.3. Multiplication of floating point numbers F_1 (with sign s_1, exponent e_1 and significand p_1) and F_2 (with sign s_2, exponent e_2 and significand p_2) is a five step process. Its flow chart is presented in Fig 2.3 [2].

Table 2.3: Floating point multiplication algorithm

Step 1

Calculate the tentative exponent of the product by adding the biased exponents of the two numbers, subtracting the bias. The bias is 127 and 1023 for single precision and double precision IEEE data format respectively

$e_1 + e_2 - bias$

Step 2

If the sign of two floating point numbers are the same, set the sign of product to '+', else set it to '-'.

Step 3

*Multiply the two significands. For p bit significand the product is 2p bits wide (p, the width of significand data field, is including the leading hidden bit (1)). Product of significands falls within rang*e:

$1 \le product < 4$

Step 4

Normalize the product if MSB of the product is 1 (i.e. product of two significands), by shifting the product right by 1 bit position and incrementing the tentative exponent.

$significands \ge 2$ *Evaluate exception conditions, if any.*

Step 5

Round the product if R(M0 + S) is true, where M0 and R represent the pth and (p+1)st bits from the left end of normalized product and Sticky bit (S) is the logical OR of all the bits towards the right of R bit. If the rounding condition is true, a 1 is added at the pth bit (from the left side) of the normalized product. If all p MSBs of the normalized product are 1's, rounding can generate a carry-out. In that case normalization (step 4) has to be done again.

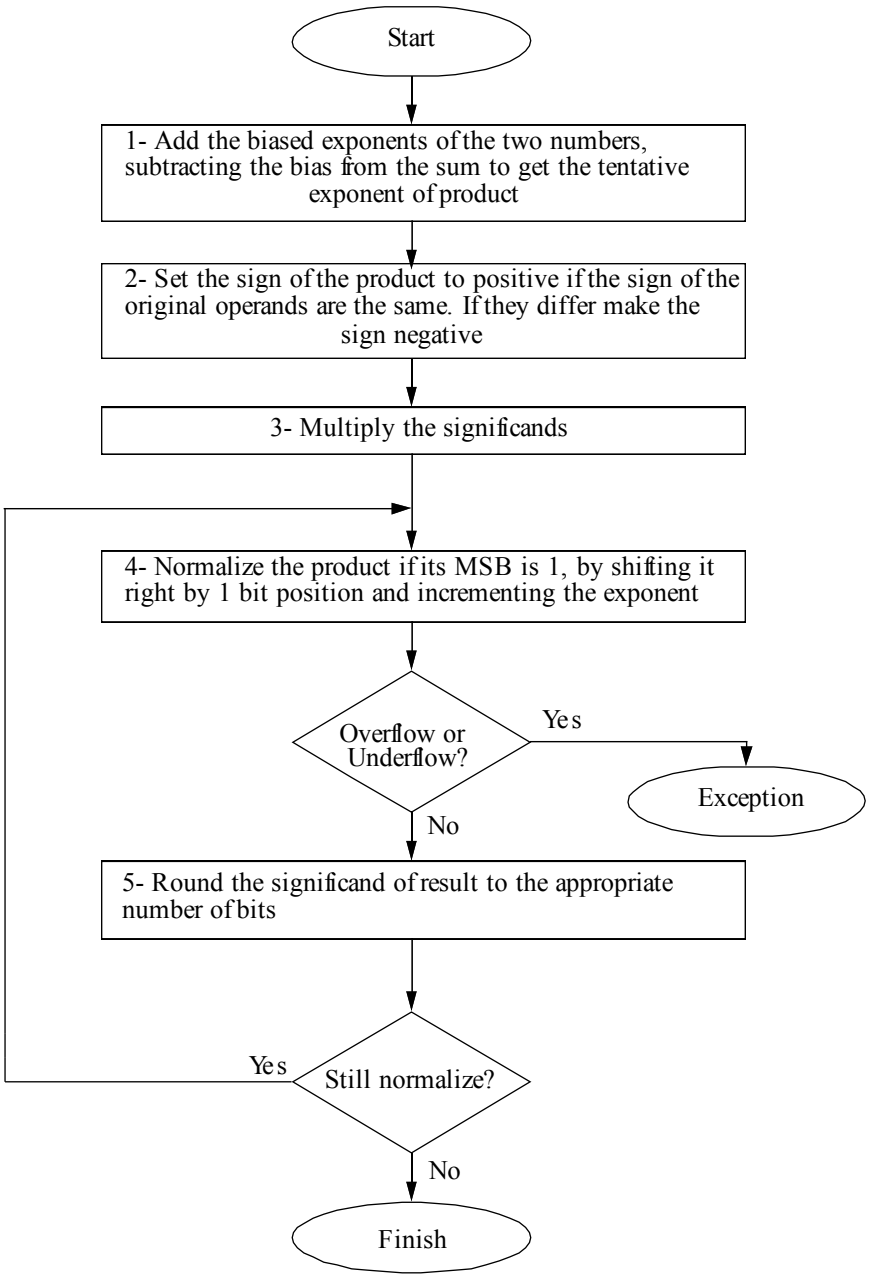


Fig 2.3 - Block diagram of IEEE compliant floating point multiplication

Fig 2.4 illustrate the process of significand multiplication, normalization and rounding.

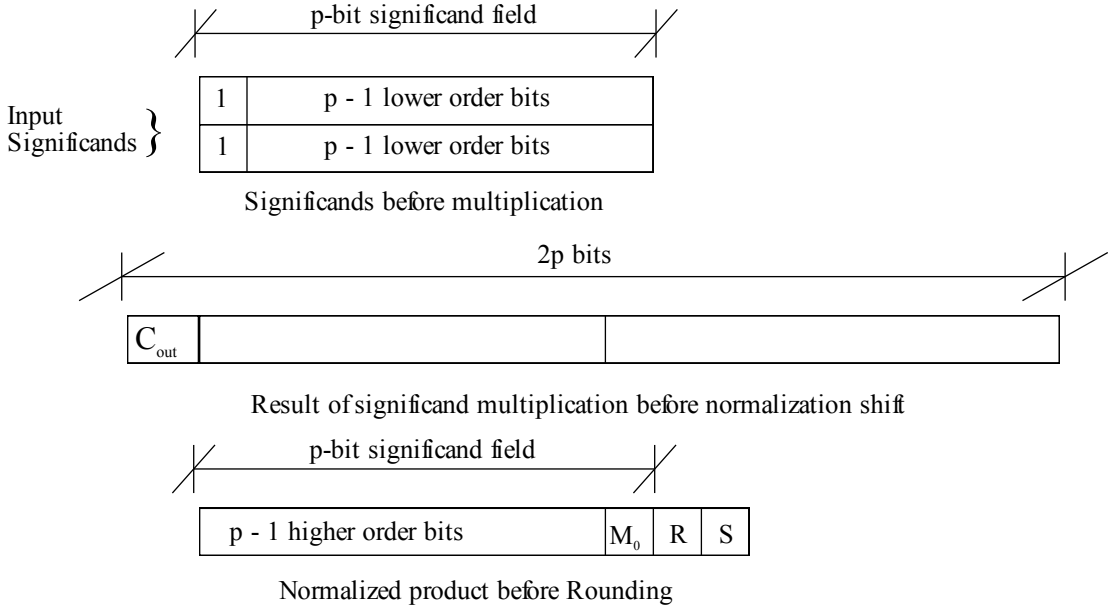


Figure 2.4 - Significand multiplication, normalization and rounding

Example:

Multiply the following two numbers. Use IEEE 754 standard:
$A= 25.5_{10}$ $B= -0.375_{10}$

Answer:
A can be represented as $A= 1.10011 * 2^4$ or $exp= 127+ 4 = 131_{10}$, Sig =1.10011, S=0

0	1 0 0 0 0 0 1 1	1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

B can be represented as $B= 1.1 * 2^{-2}$ or $exp= 127 -2= 125_{10}$, sig =1.1 , S=1

1	0 1 1 1 1 1 0 1	1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Add exponent and subtract bias
Exponent = 1 0 0 0 0 0 1 1 + 0 1 1 1 1 1 0 1 - 0 1 1 1 1 1 1 1 = 1 0 0 0 0 0 0 1

Multiply Significands 1.1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 *
1.1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

10.01100100000000000000000000000000000000000000

Now round the results.

After rounding we get: 10. 01100100000000000000000
After normalization we get: 1. 0 01100100000000000000000 * 2^1

New exponent after normalization 1 0 0 0 0 0 0 0 1 + 0 0 0 0 0 0 0 1 = 1 0 0 0 0 0 1 0
Final result

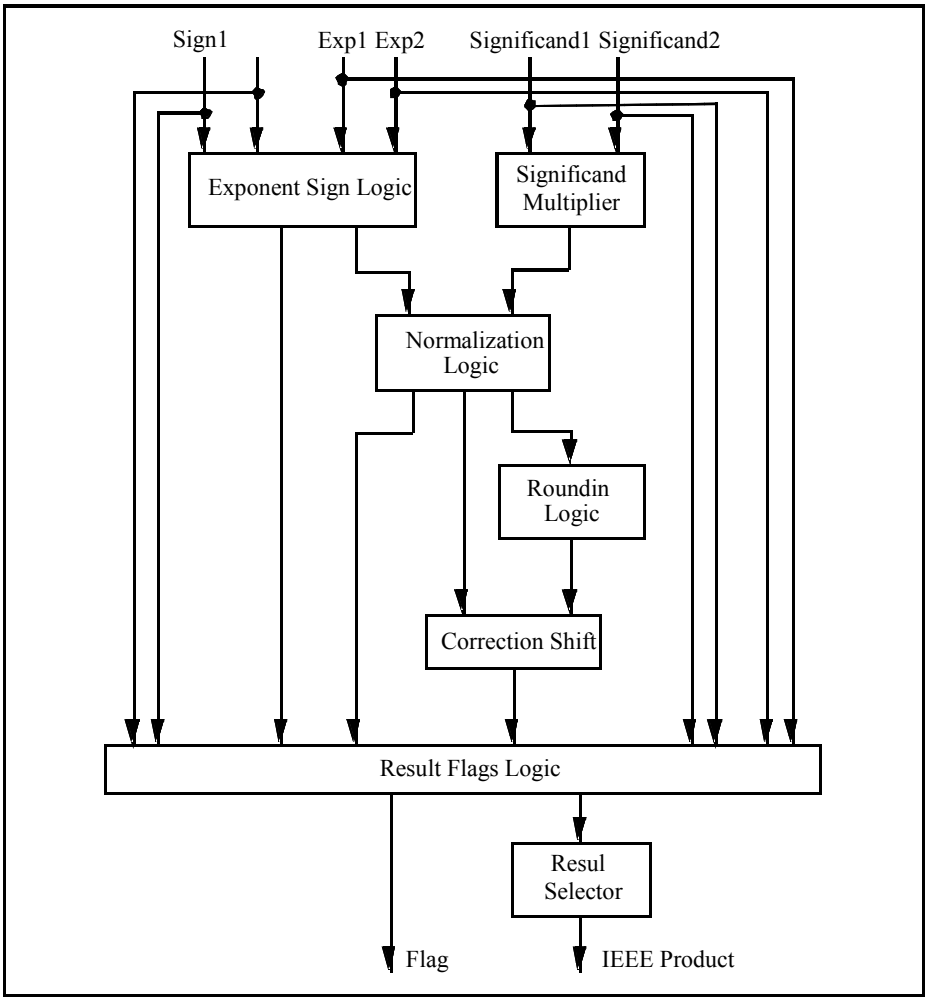
1	1 0 0 0 0 0 1 0	0 0110 010 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Unbiased value of exponent is 1000 0010 -0111 1111 = 0000-0011 ie $(130-127 = 3)_{10}$

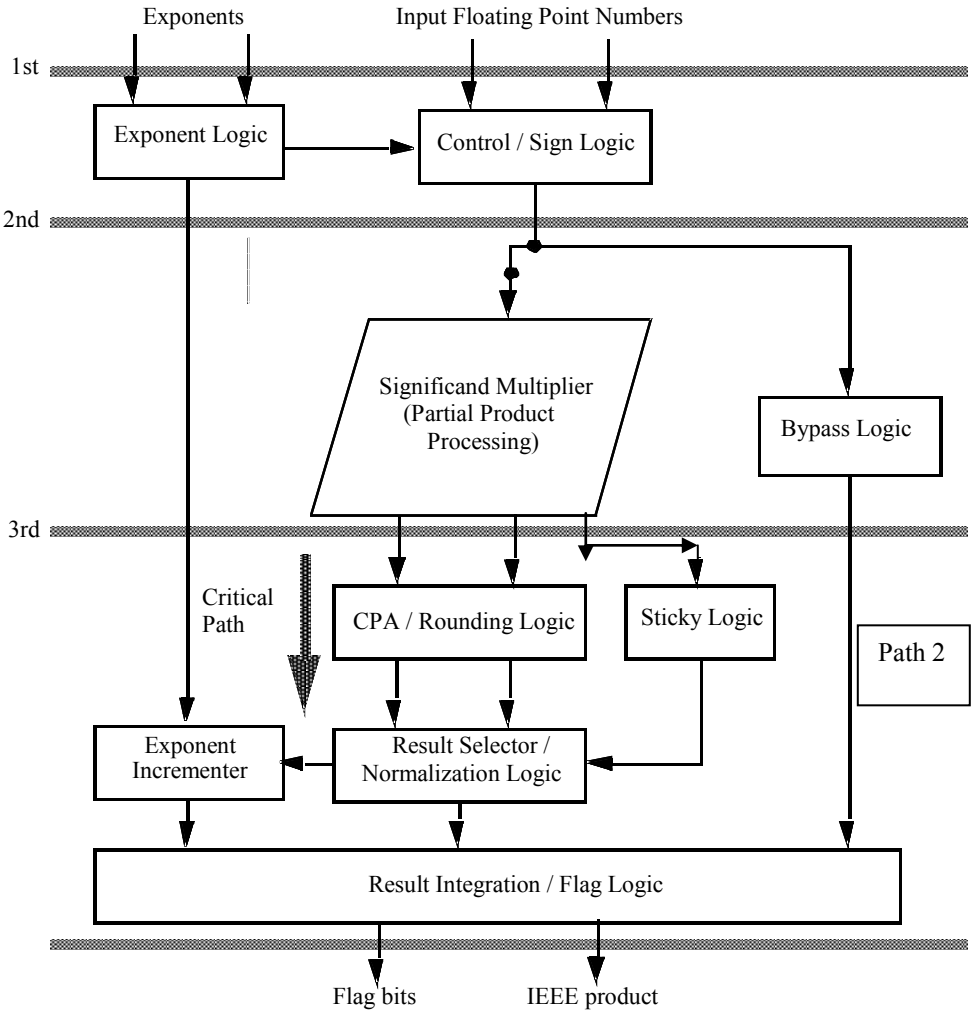
A * B = 1. 0011001 * 2^3 = -9.5625_{10}

Multiplier Architecture

A simple multiplier architecture is shown below:
The exponent logic is responsible for extracting the exponents and adding them and subtracting the bias. The Control/Sign logic, decodes the sign bit (EXOR) and directs the significands to the integer multiplier.
The results of the significands multiplication is rounded by the rounding logic and if necessary is normalized through a shift operation. The exponent in updated by the exponent increment block and the final results, are presented to the output logic. This architecture is shown in figure below



This architecture can be improved by addition of two features as shown in the diagrams below. A bypass is added so that Not-A-Number, such as non computing operations can bypass the multiplication process. The architecture also features pipelining lines, where the multiplier can be made to operate faster at the expense of latency.



Comparaison for the Scalable Single Data Path FPM, Double Data Path FPM and Pipelined Double Data Path FPM is also done for IEEE single precision data format in order to validate the findings that DDFPM require less power as compared to SDDFPM. Table 2.4 below shows the results obtained by synthesizing the three design using 0.22 micron CMOS technology.

Table 2.4 : Comparaison of SDFPM,DDFPM,PDDFPM

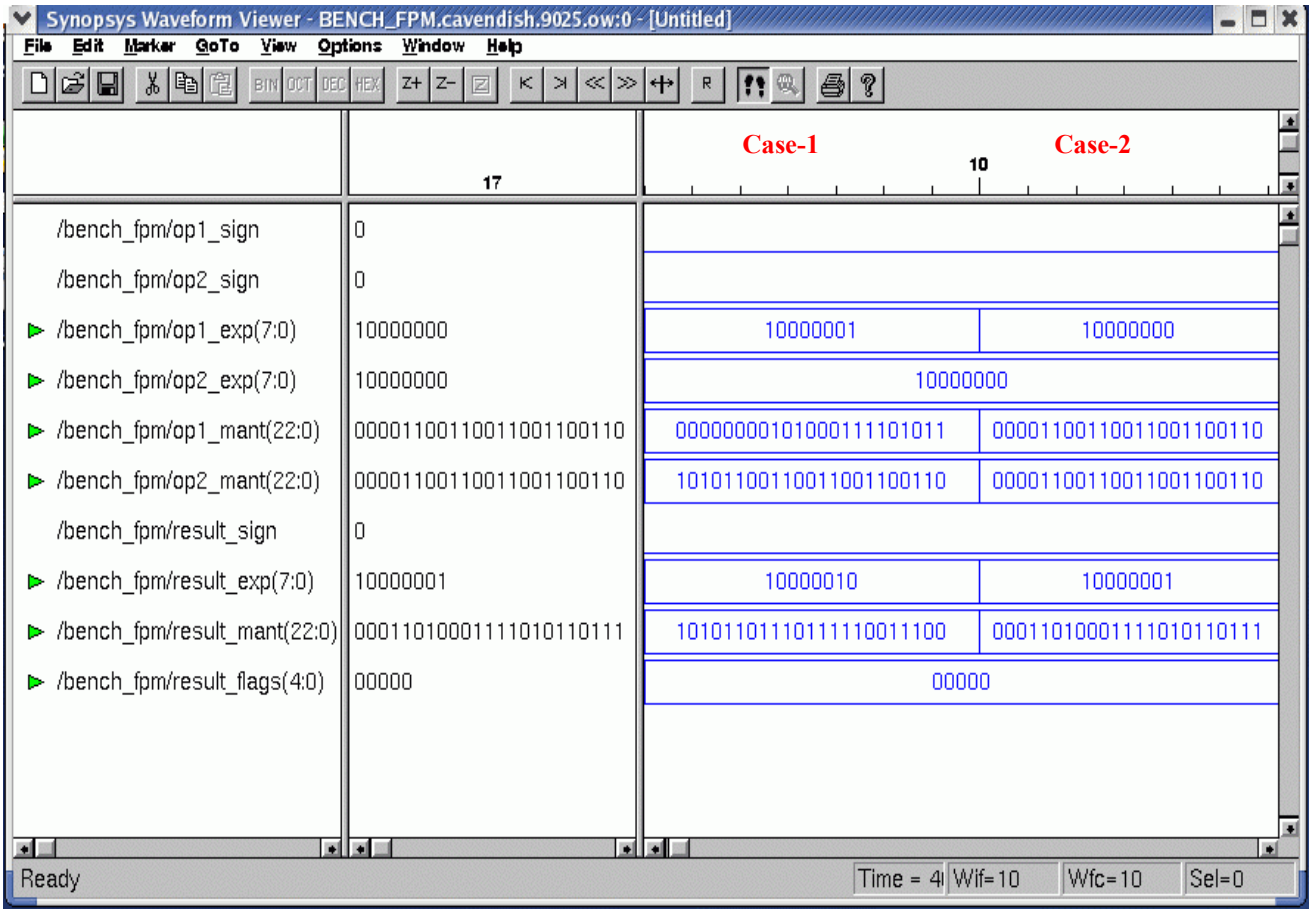
	AREA (cell)	**POWER** (mW)	**Delay** (ns)
Single Data Path FPM	2288.5	204.5	69.2
Double Data Path FPM	2997	94.5	68.81
Pipelined Double Data Path FPM	3173	105	42.26

As we can see from the Table 2.4 that the power reduction is quite significant in case of a DDFPM as compare to SDFPM which is almost 53% . This validate our findings for the the DDFPM require less power.

Pipelined DDFPM is designed in order to reduce the overall delay without much increase in area and power. The findings show that the worst case delay is reduced by almost 39%, however there is 5.5% increase in area and 10% increase in power which is quite acceptable.

Table Test Cases for IEEE Single Precision for SDFPM

		S	Exponent	Significand
Case-1 Normal Number	Operand1	0	10000001	00000000101000111101011
	Operand2	0	10000000	10101100110011001100110
	Result	0	10000010	10101101110111110011100
Case-2 Normal Number		**S**	**Exponent**	**Significand**
	Operand1	0	10000000	00001100110011001100110
	Operand2	0	10000000	00001100110011001100110
	Result	0	10000001	00011010001111010110111



Above is the synopsys simulation results of Single Data Path FP Multiplier

2.5 Floating Point Addition

The algorithm of addition of floating point numbers F_1 (with sign s_1, exponent e_1 and significand p_1) and F_2 (with sign s_2, exponent e_2 and significand p_2) is listed in Table 2.5 [1], and block diagram is presented in Fig 2.5 [2].

Table 2.5: Floating point addition algorithm

Step 1 *Compare the exponents of two numbers for* and calculate the absolute value of difference between the two exponents . Take the larger exponent as the tentative exponent of the result. $e_1 > e_2$ $e_1 \leq e_2$ Step 2 $\lvert e_1 - e_2 \rvert$ *Shift the significand of the number with the smaller exponent right through a number of bit positions that is equal to the exponent difference. Two of the shifted out bits of the aligned significand are retained as guard (G) and Round (R) bits. So for p-bit significands, the effective width of aligned significand must be p + 2 bits. Append a third bit, namely the sticky bit (S), at the right end of the aligned significand. The sticky bit is the logical OR of all shifted out bits.* Step 3 *Add/subtract the two signed-magnitude significands using a (p + 3)-bit adder. Let the result of this is SUM.* Step 4 *Check SUM for carry out (C_{out}) from the MSB position during addition. Shift SUM right by one bit position if a carry out is detected and increment the tentative exponent by 1. During subtraction, check SUM for leading zeros. Shift SUM left until the MSB of the shifted result is a 1. Subtract the leading zero count from tentative exponent.* *Evaluate exception conditions, if any.* Step 5 *Round the result if the logical condition R''(M_0 + S'') is true, where M_0 and R'' represent the pth and (p + 1)st bits from the left end of the normalized significand. New sticky bit (S'') is the logical OR of all bits towards the right of the R'' bit. If the rounding condition is true, a 1 is added at the pth bit (from the left side) of the normalized significand.* *If p MSBs of the normalized significand are 1's, rounding can generate a carry-out. In that case normalization (step 4) has to be done again.*

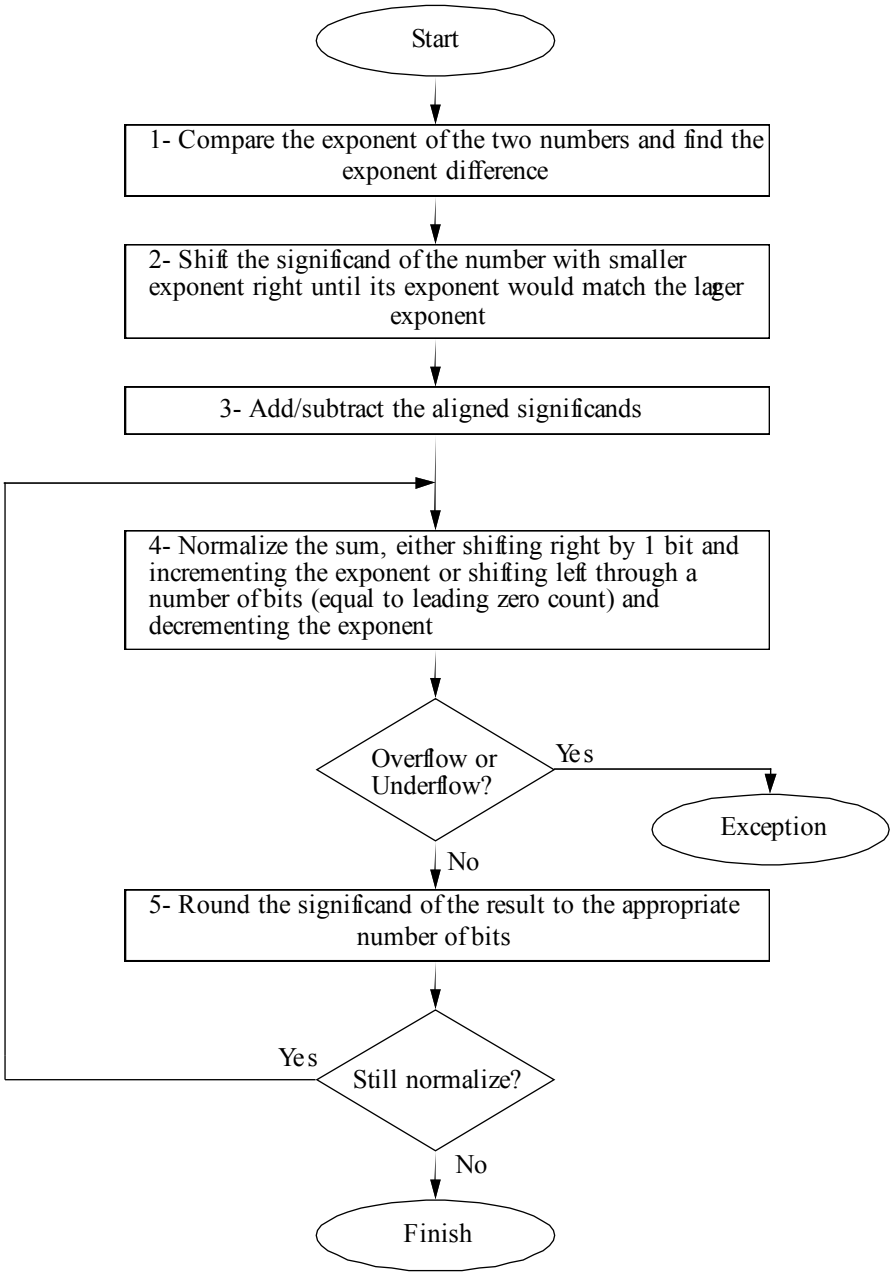


Fig 2.5 - Block diagram of IEEE compliant floating point addition

Example 2

Add the following two numbers. Use IEEE 754 standard:
$A= 25.5_{10}$ $B= -63.25_{10}$

Answer:
A Can be represented as:
$A= 1.10011 * 2^4$ or exp= $127+ 4 = 131_{10}$, Sig =1.10011, S=0

0	1 0 0 0 0 0 1 1	1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

B Can be represented as
$B= 1.1111101 * 2^5$ or exp= $127 +5= 132_{10}$, Sig =1. 1111101 , S=1

1	1 0 0 0 0 1 0 0	1 1 1 1 1 0 10 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Compare the exponents and determine the bigger number and make it the reference.
10 0 0 0 1 0 0 - 1 0 0 0 0 0 1 1 = 1

Shift the smaller number to the right by one place (normalizing to the exponents difference of 1) gives significance of $A= 0.1100111 * 2^5$

Now Add both numbers together

```
A    0.1100110 +
B    1.1111101
-----------------------------------
```

Since B is –ve then taking it 2's complement and performing addition, we get

```
A    00.1100110   +
B    10.0000011   2's Complement of B
-----------------------------------
      10.1101001
```

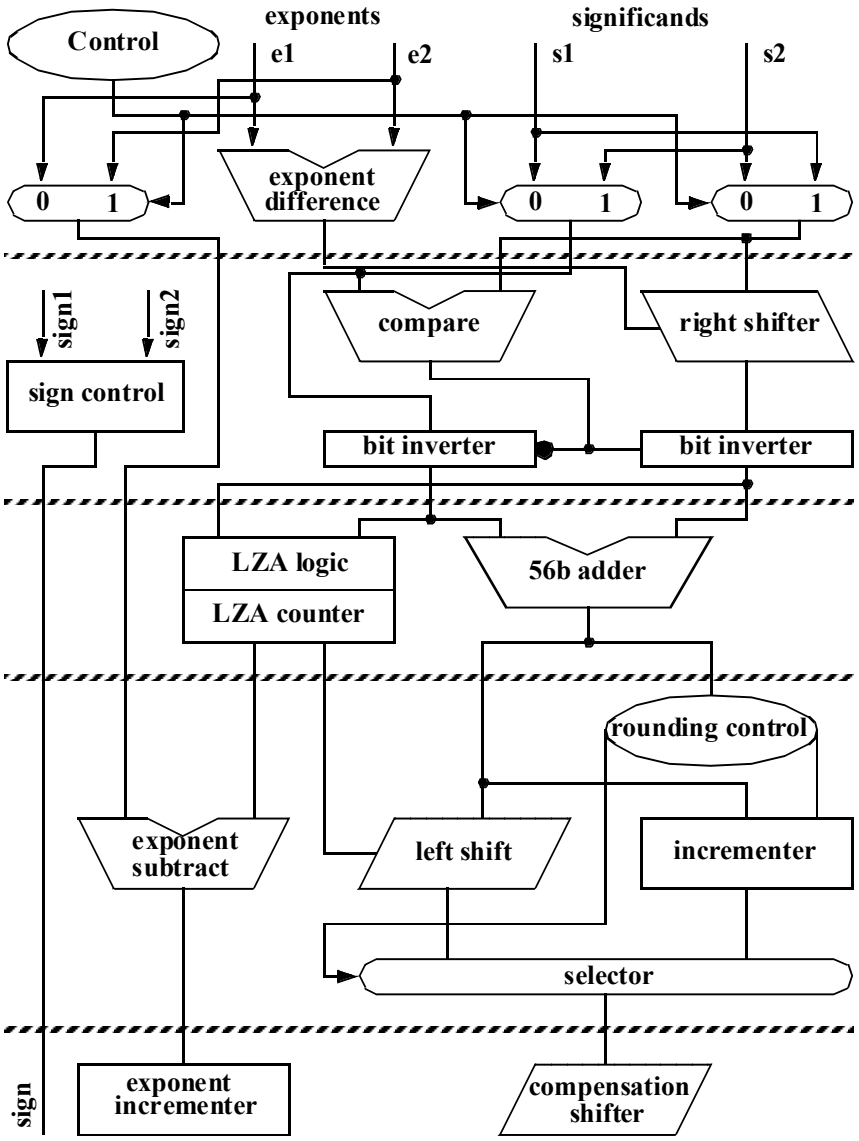
Which is a –ve number. Taking its sign and magnitude gives the results as Significand of the result = 0 1 . 0 0 1 0 1 1 1. with S=1
Therefore the results can now be integrated as

1	1 0 0 0 0 1 0 0	0 0 1 0 1 1 10 0 0 0 0 0 0 0 0 0 0 0 0 0 0

This is equal to $- 1. 0 0 1 0 1 1 1 * 2^5 = - 37.75_{10}$

Floating Point Adder Architecture

A block diagram of floating Point adder is shown below. Exponents, sign bits and the significands are fed into the adder. The exponents subtractor gives the amount of shift, while the comparator gives which operands is to be shifted. The right shift of the smaller number is achieved by a barrel shifter. Inversion of one operand is achieved by the sign bits and the bit inverters. The Adder adds the operands and passes the results to the rounding logic. Leading Zero Anticipator logic determines the normalization needed where the results are normalized and the exponents are adjusted. Finally the right significand is selected and is passed to the output together with the exponents and the sign. This architecture features two additional non standard blocks The LZA logic and the lines where pipelining registers can be inserted to speed up the design



2.7 Exceptions

For the handling of arithmetic operations on floating point data, IEEE standard specifies some exception flags. Some exception conditions are listed in Table 2.5 [3]. When the results of an arithmetic operation exceeds the normal range of floating point numbers as shown in Fig 2.7 [2], overflow or underflow exceptions are initiated. Please see Table 2.6

Table 2.5: Exceptions in IEEE 754

Exception	Remarks
Overflow	Result can be ± ∞ or default maximum value
Underflow	Result can be 0 or denormal
Divide by Zero	Result can be ± ∞
Invalid	Result is NaN
Inexact	System specified rounding may be required

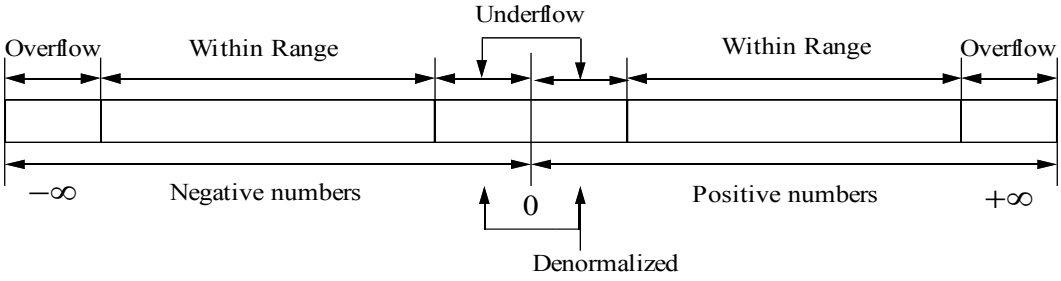


Fig 2.7 - Range of floating point numbers

Table 2.6: Operations that can generate Invalid Results

Operation	Remarks
Addition/ Subtraction	An operation of the type ∞ ± ∞
Multiplication	An operation of the type 0 x ∞
Division	Operations of the type 0/0 and ∞/∞
Remainder	Operations of the type x REM 0 and ∞ REM y
Square Root	Square Root of a negative number

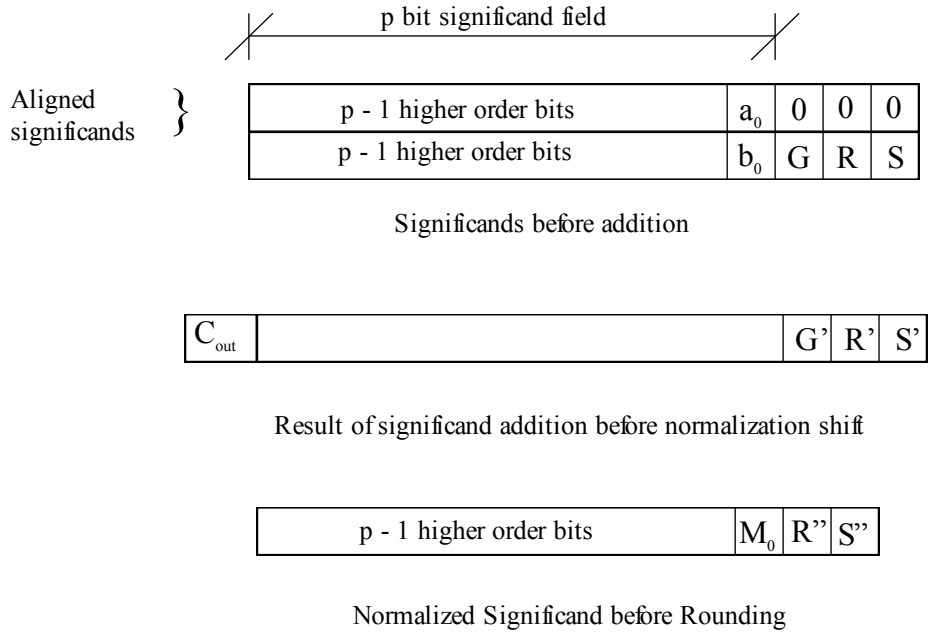


Fig 2.6 - Significand addition, normalization and rounding

Detection of overflow or underflow is quite straight forward as the range of floating point numbers is associated with the value of exponents. Table 2.6 [1] lists all possible operations that result in an invalid exception. During invalid exception the result is set to a NaN (not a number). Inexact exceptions are true whenever the result of a floating point operation is not exact and IEEE rounding is required [1]. In IEEE single precision data format, width of exponent field is 8, so 256 combinations of exponent are possible. Among them two are reserved for special values. The value $e = 0$ is reserved to represent zero (with fraction $f = 0$) and denormalized numbers (with fraction). The value $e = 255$ is reserved for (with fraction $f = 0$) and NaN (with fraction). The leading bit of significands (hidden bit) is zero instead of one for all the special quantities.

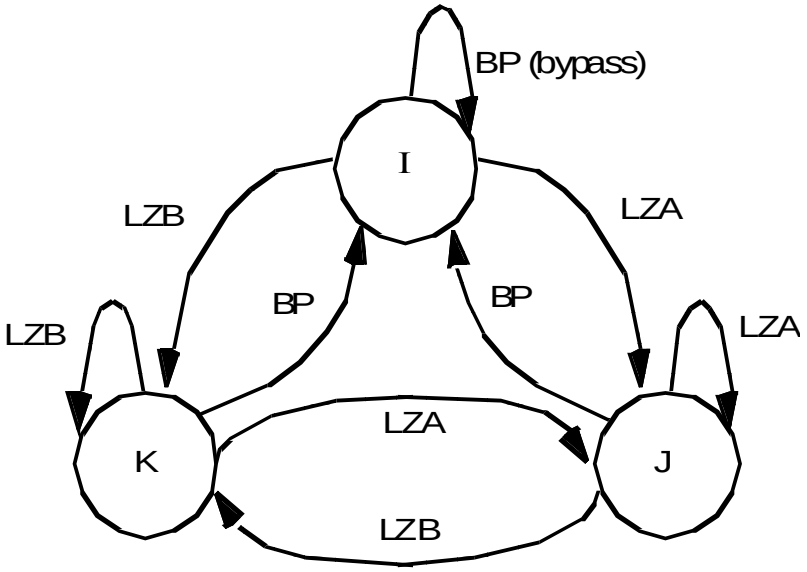
$f \neq 0$

$\pm\alpha$

$f \neq 0$

Low Power Triple Data Path Adder, TFADD

In literature several architectures exist that improves on the basic adder. The TFADD is such an architecture. It uses the range of values of the input data to decide on which path the data should be processed. This architecture provides 3 data paths for data processing with the bypass path being a non computing path. Data that do not require calculation such as addition with zero or NaN passes through this path. The architecture is shown in the figure below with the control state machine. If the exponents difference between the two operands is "1" then the then there is no need for initial shift of one operand thus a barrel shifter can be eliminated from this path, however we will need a barrel shifter for normalization. On the other hand if the exponent difference is greater than 1, then there is a barrel shifter for initial shifting of one operand, but there will be no need for a barrel shifter for normalization. Thus each path is shorter by one barrel shifter. Depending on the operands value and sign, this method has less delay and less power at the expense of extra area. The finite state machine below shows the operation of the adder. State I is the bypass state, while state K, requires no normalization barrel shifter. State J the data follows the middle path where there is no need for the exponent alignment barrel shifter.



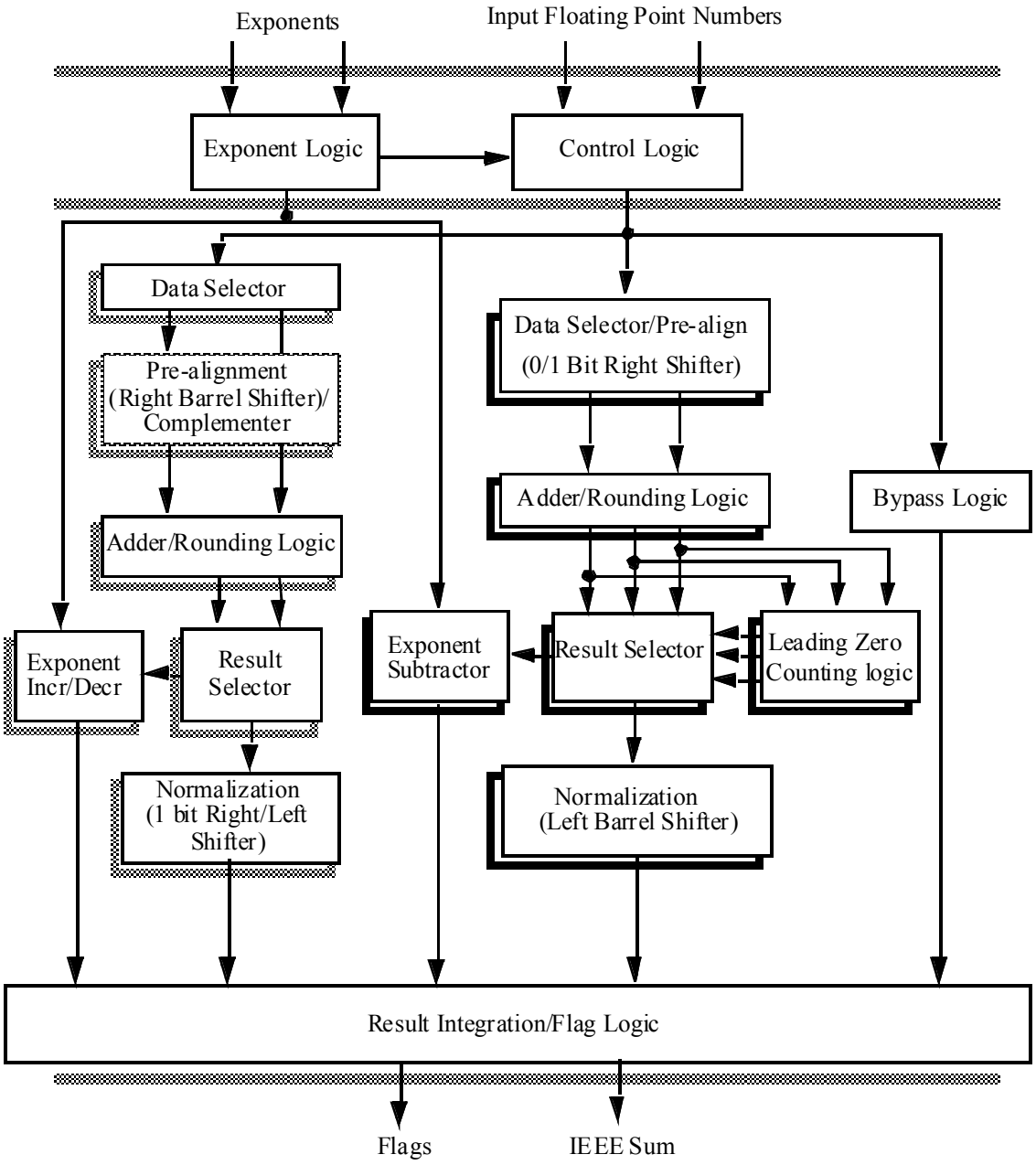
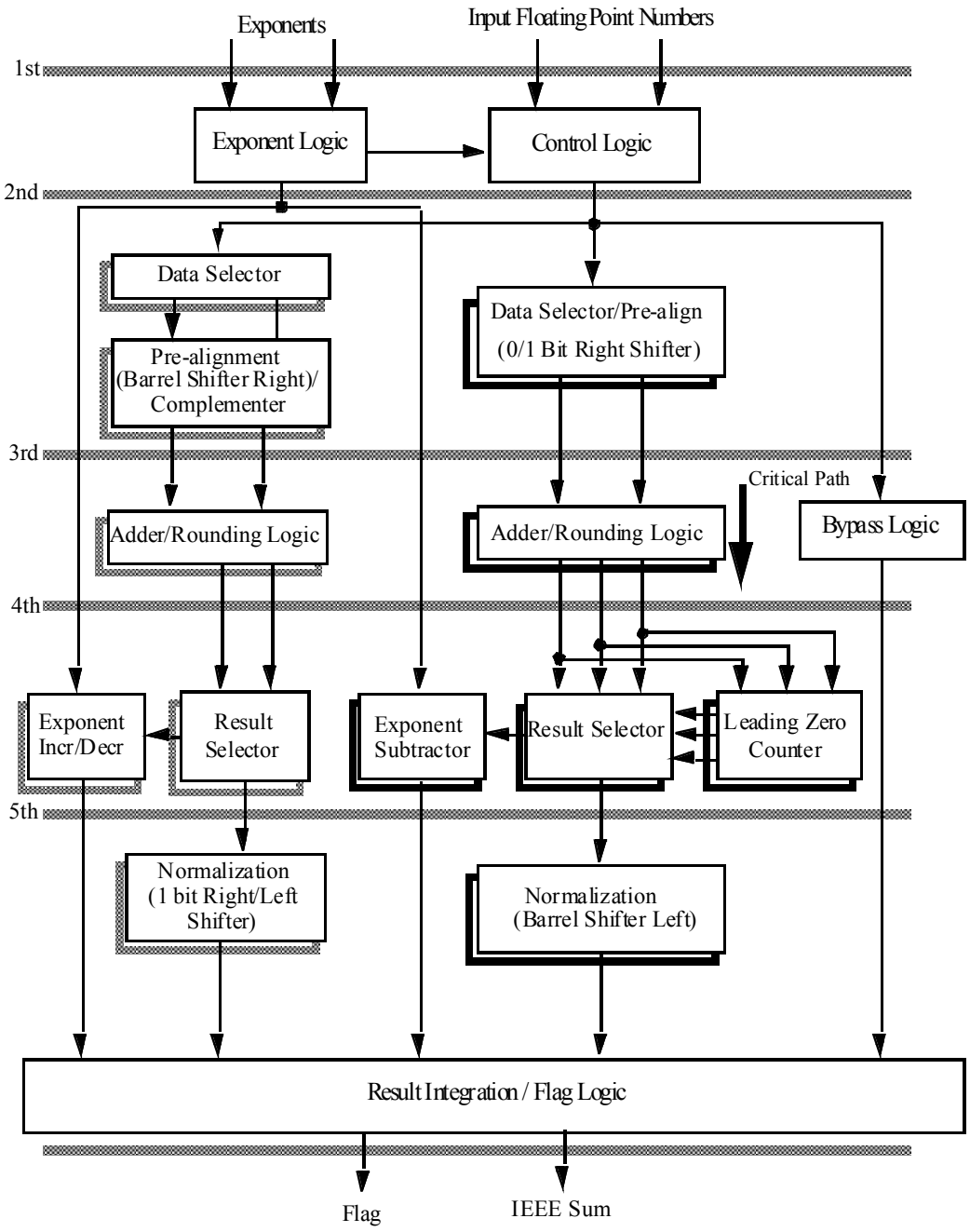


Fig 4.2 - Block diagram of the TDPFADD



This figure shows pipelining the TPFADD to speed up its operation

TEST RESULTS [21]

Several tests on data has been carried out. In particular IEEE standard 754 for the single and double precision have been tested for a variety of inputs to see its performance in extreme conditions. The code out of the generator has also been synthesized by Synopsys using Xilinx 4052XL-1 FPGA technology. The results are shown in Table below.

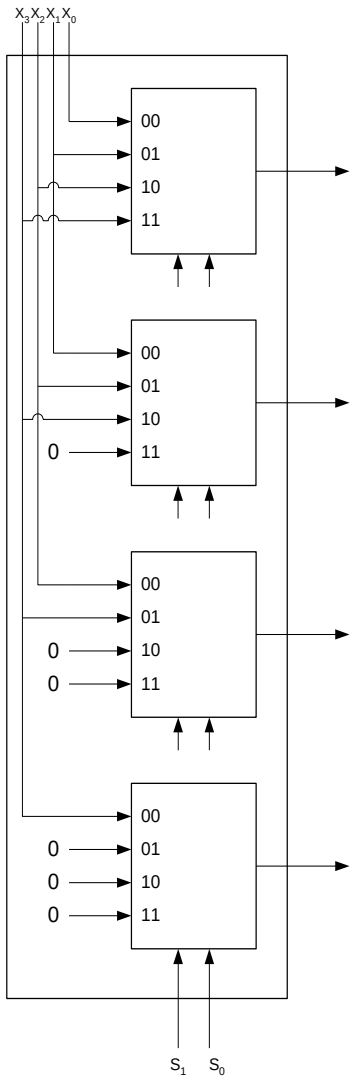
Table Comparison of Synthesis results for IEEE 754 Single Precision FP addition. Using Xilinx 4052XL-1 FPGA

PARAMETERS	SIMPLE	TDPFADD	PIPE/ TDPFADD
Maximum delay, D (ns)	327.6	213.8	101.11
Average Power, P (mW)@ 2.38 MHz	1836	1024	382.4
Area A, Total number of CLBs (#)	664	1035	1324
Power Delay Product (ns. 10mW)	7.7. $*10^4$	4.31 $*10^4$.	3.82 $*10^4$
Area Delay Product (10 # .ns)	2.18`$*10^4$	2.21 * 10^4	1.34 $*10^4$
Area-Delay^2 Product (10# . ns^2)	7.13.$*10^6$	4.73 * 10^6	1.35 $*10^6$

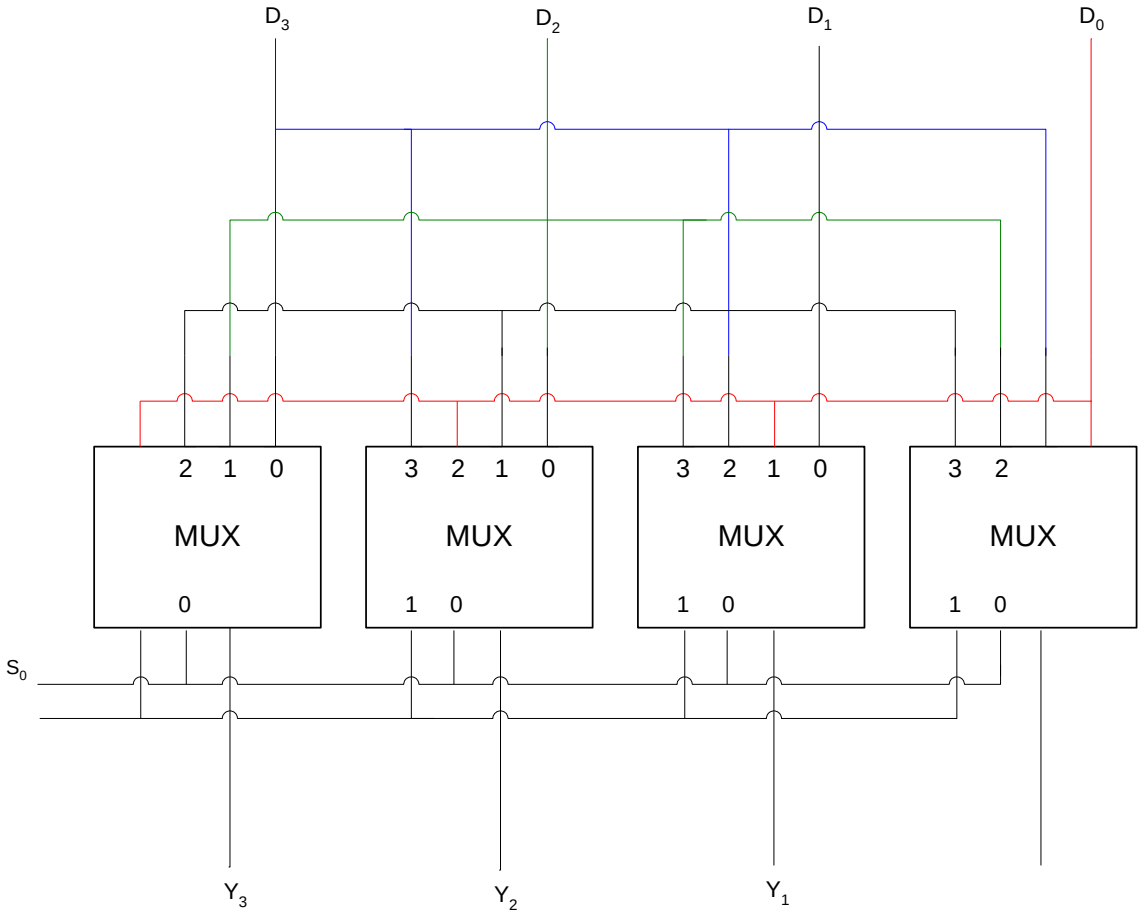
Barrel Shifters

In many applications such as floating point adders or multipliers circuits are needed that can shift several data items in one move, such a circuit is named barrel shifter. A variety of barrel shifters exist each targeted towards a special application. We will not cover all

applications rather the principal operation. The figure below shows a right shift barrel shifter constructed from four 4-1 multiplexer that performs 0, 1, 2 ,3 bits right shift of data $x_3\ x_2\ x_1\ x_0$ in response to the control input $S_1\ S_0$.



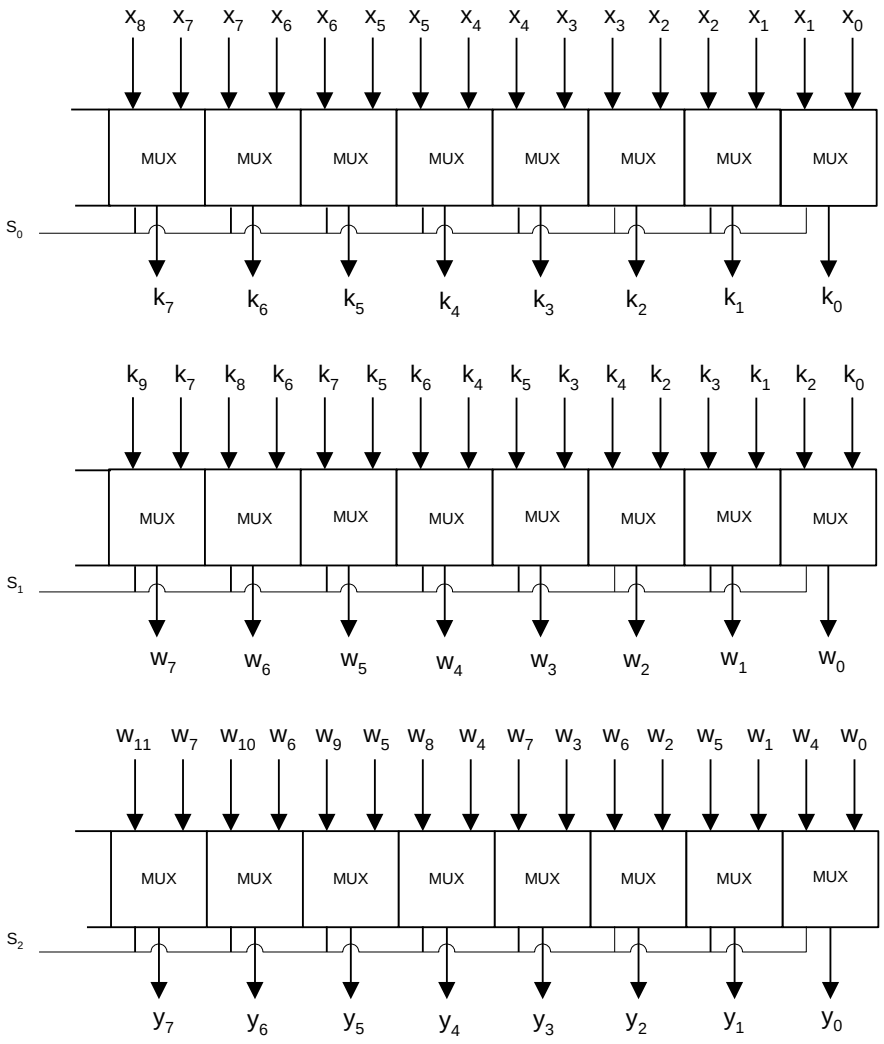
For example if S_1S_0 is 11 then the output is $000x_3$. It is possible to change the shift order by reconfiguring the inputs. It is also possible to make the data rotate, by appropriate connection of input of the multiplexer. The circuit below shows a barrel shifter that performs rotation of data input in accordance with the control table given below.



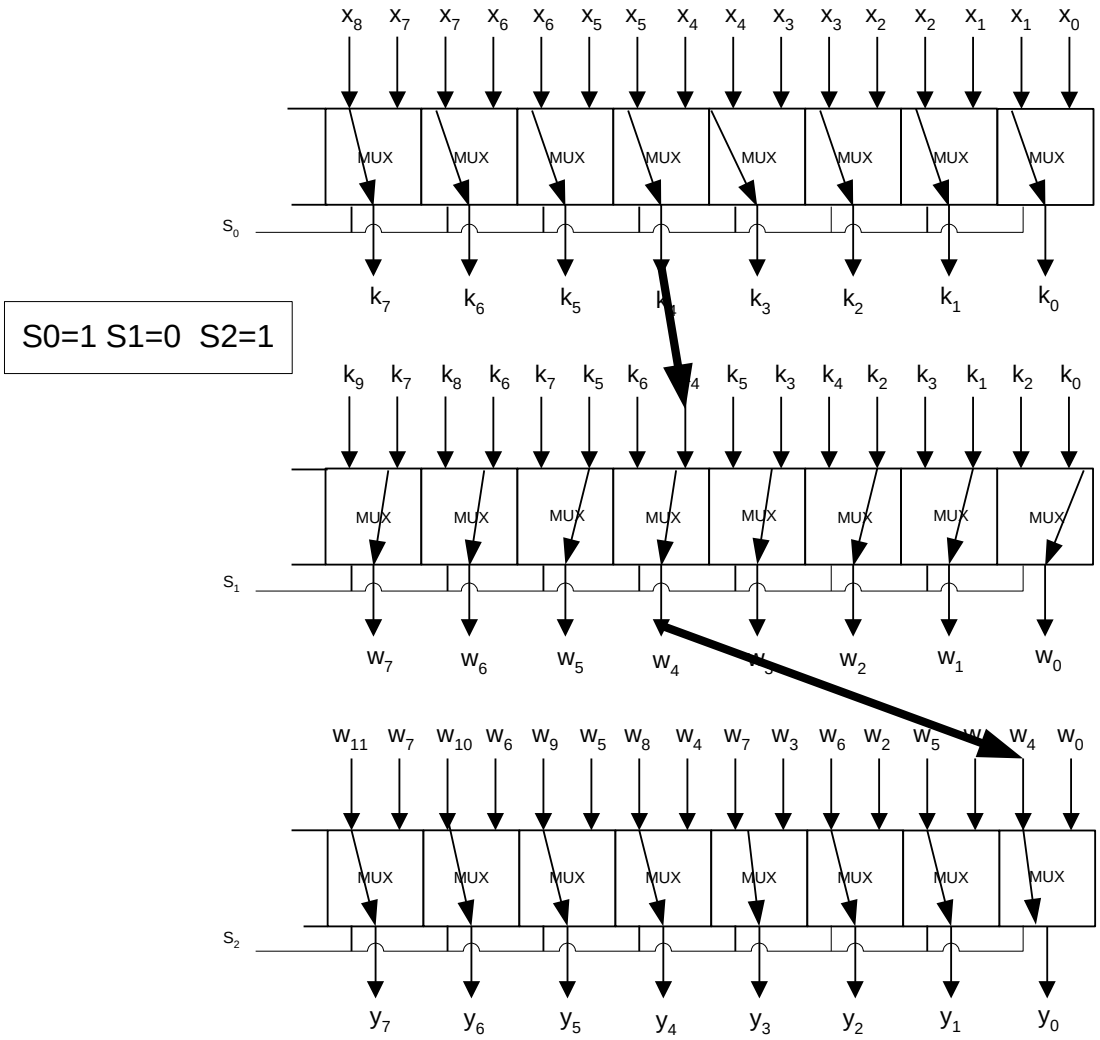
Select		Out Put				Operation
S_i	S_o	Y_3	Y_2	Y_1	Y_0	
0	0	D_3	D_2	D_1	D_0	No Shift
0	1	D_2	D_1	D_0	D_3	Rotate Once
1	0	D_1	D_0	D_3	D_2	Rotate Twice
1	1	D_0	D_3	D_2	D_1	Rotate 3 times

Most barrel shifters however are designed with 2 to1 MUXs using distributed shifting method. With the distributed method the delay is always proportional to $\log_2 n$ where n is the number of shifts required.

The figure below, shows this principal. At the first level of 2-1 Mux data are connected into the MUX with one data difference. The out from this MUX is connected to a second level of 2-1 MUX, with data connected with a difference of two bits. With the 3rd level the process is repeated with 4 bits of shift.



The diagram below shows setting S2=1, S1=0, S0=1, which gives us 5 shifts of data to the right. The path of the first data bit output is shown.



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Appendix [20]

Introduction to IEEE-754 standard

In the early days of computers, vendors start developing their own representations and methods of calculations. These different approaches lead to different results in calculations. So the IEEE organization defined in the IEEE-754 standard a representation of the floating point numbers and the operations.

Representation

As in all floating point representations, the IEEE representation divides the number of bits into three groups, the sign, the exponent and the fractional part.

Fractional part

Fractional part is represented as sign-magnitude, which needs a reserved bit for the sign.

The exponent

The exponent is based on the biased representation. This means if k is the value of the exponent bits, then the exponent of the floating-point number is k – the bias. So to represent the exponent *zero* the bits should hold the value of the bias.

A.1 Hidden-bit

Another feature of the IEEE representation is the hidden bit. This bit is the only bit to the left of the fraction point. This bit is assumed to be 1, which gives an extra bit of storage in the representation to increases the precision.

Sign Bit

The sign bit is as simple as it gets. 0 denotes a positive number; 1 denotes a negative number. Flipping the value of this bit flips the sign of the number.

A.2 Precision

The IEEE-754 defines set of precisions which depends on the number of bits used. There are two main precisions, the single and the double.

A.2.1 Single Precision

The IEEE single precision floating point standard representation require a 32 bit word, which may be represented as numbered from 0 to 31, right to left as shown

A.2.2 Double Precision

The IEEE single precision floating point standard representation require a 32 bit word, which may be represented as numbered from 0 to 63, right to left as shown

Table A.1: Exponent range and number of bits in single and double precision floating-point representation.

	Single	Double
Exponent(max)	+127	+1023
Exponent(min)	-126	-1022
Exponent Bias	+127	+1023
Precision (#bits)	24	53
Total Bits	32	64
Sign bits	1	1
Exp Bits	8	11
Fraction	23	52

A.3 Normalization

Normalization is the act of shifting the fractional part in order to make the left bit of the fractional point "1". During this shift the exponent is incremented.

A.3.1 Normalized numbers

Normalized numbers are numbers that have their MSB a "1" in the most left bit of the fractional part.

A.3.2 Denormalized numbers

Denormslized numbers are the opposite of the normalized numbers. (i.e. the MSB 1 is not in the most left bit of the fractional part).

Operations:

Some operations require that the exponent field be the same for all operands (like addition). In this case one of the operands should be denormalized.

A.3.3 Gradual underflow:

One of the advantages of the denormalized numbers is the gradual underflow. This came from the fact the normalized number that can represent minimum number is 1.0×2^{min} and all numbers smaller than that are rounded to zero (which means there are no numbers between 1.0×2^{min} and 0. The denormalized numbers expands the range and gives gradual underflow through the division of the range between 1.0×2^{min} to 0 with the same steps as the normalized numbers.

A.4 Special values

The IEEE-754 standard supports some special values that gives special functions and give some signals.

Table A.2: Special values

Name	Exponent	Fraction	sign	Exp Bits	Fract Bits
+0	min−1	= 0	+	All zeros	All Zeros
−0	min−1	= 0	−	All zeros	All Zeros
Number	min ≤ e ≤ max	any	any	Any	Any
+∞	max+1	= 0	+	All ones	All zeros
−∞	max+1	= 0	−	All ones	All zeros
NaN	Max+1	≠ 0	any	All ones	Any

A.4.1 Zero

The zero is represented as a signed zero (-0 and +0)

It is represented as min□-1 in the exponent and zero in the fraction.

The signed zero is important for operations that preserve the sign like multiplication and division. It is also important to generate +∞ or -∞.

A.4.2 NaN

Some computations generate undefined results like 0/0 and √[(-1)]. These operations should be handled or we will get strange results and behavior. *NaN* is defined to be generated upon these operations and so the operations are defined for it to let the computations continue.

Whenever a NaN participates in any operation the result is NaN.

There is a family of NaN according to the above table and so the Implementations are free to put any information in the fraction part.

All comparison operators (= , < , ≤ , > , ≥) (except (≠)should return false when NaN is one of its operands.

Table A.3: Sources of NaN

Operation	Produced by
+	∞+(−∞)
×	0×∞
/	0/0, ∞/∞

A.4.3 Infinity

The infinity is like the NaN, it is a way to continue the computation when some operations are occurred.

Generation:
Infinity is generated upon operations like x/0 where $x \neq 0$

Results:
The results of operations that get ∞ as a parameter is defined as: "Replace the ∞ by the limit $\lim_{x \to \infty}$. For example $3/\infty = 0$ because $\lim_{x \to \infty} 3/x = 0$ and $\sqrt{\{\infty\}} = \infty$ and $4-\infty = -\infty$

A.5 Exceptions

Exceptions are important factors in the standard to signal the system about some operations and results.

When an exception occurs, the following action should be taken:

- A status flag is set.
- The implementation should provide the users with a way to read and write the status flags.
- The Flags are ``**sticky**'' which means once a flag is set it remains until its explicitly cleared.
- The implementation should give the ability to install trap handlers that can be called upon exceptions.

Common exceptions in floating-point numbers are:

- Overflow, underflow and division by zero:

  As is obvious from the table below, the distinction between Overflow and division by zero is to give the ability to distinguish between the source of the infinity in the result.
- **Invalid:**

  This exception is generated upon operations that generate NaN results. But this is not a reversible relation (i.e. if the output is NaN because one of the inputs is NaN this exception will not raise).
- **Inexact:**

It is raised when the result is not exact because the result can not be represented in the used precision and rounding cannot give the exact result.

Table A.4: Exceptions in IEEE 754 standard

Exception	Cased by	Result
Overflow	Operation produce large number	±∞
Underflow	Operation produce small number	0
Divide by Zero	x/0	±∞
Invalid	Undefined Operations	NaN
Inexact	Not exact results	Round(x)

A.6 IEEE Rounding:

As not all real numbers can be represented precisely by floating point representation, there is no way to guarantee absolute accuracy in floating point computations. Floating point numbers are approximations of real numbers. Also the accuracy of results obtained in a floating point arithemetic unit is limited, even if the intermediate results calculated in the arithematic unit are accurate. The number of the computed digits may exceed the total number of digits allowed by the format and extra digits have to be disposed before the final results are stored in user-accessible register or memory.

IEEE 754 standard prescribes some rounding schemes to ensure acceptable accuracy of floating point computations. The standard requires that numerical operations on floating point operands produce rounded results. That's is, exact results should be computed and then rounded to the neareast floating point number using the "round to nearest – even" approach. But in practice, with limited precision hardware resources, it is impossible to compute exact results. So two guard bits (G & R) and third sticky (S) bit, are introduced to ensure the computation of results within acceptable accuracy using minimum overhead.

The default rounding mode specified by the IEEE 754 is round to nearest-even. In this mode, the results are rounded to the nearest values and in case of a tie, an even value is chosen. Table A .5 shows the operation of round to nearest – even, for different instances of significand bit patterns. In this table X represents all higher order bits of the

normalized significand beyond the LSBs that take part in rounding while the period is separating p MSBs of the normalized significand from round (R) and sticky (S) bits.

Table A.5: Round to nearest – even rounding

Significand	Rounded Result	Error	Significand	Rounded Result	Error
X0.00	X0.	0	X1.00	X1.	0
X0.01	X0.	- 1/4	X1.01	X1.	- 1/4
X0.10	X0.	- 1/2	X1.10	X1. + 1	+ 1/2
X0.11	X1.	+ 1/4	X1.11	X1. + 1	+ 1/4

It can be seen from the table that the average bias (which is the average of the sum of errors for all cases) for the round to nearest scheme is zero. Fig A.1illustrate the relative positions of the decision making bits. Rounding to the nearest value necessitate a conditional addition of 1/2 ulp (units in the last place). The decision for such addition can be reached through the evaluation of the LSB (M_0) of the most significand p bits of the normalized significand, the round (R) bit and the sticky (S) bit. Rounding is done only if R(M_0 + S) condition is true.

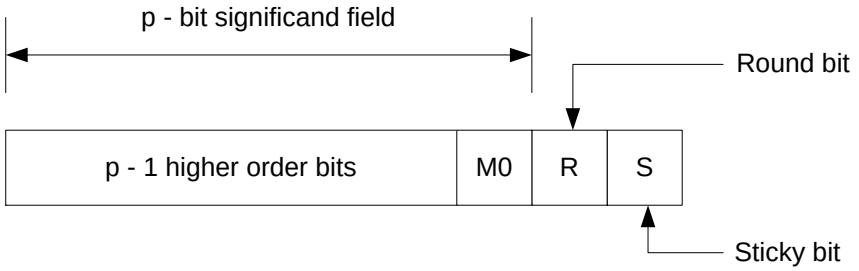

Figure A.1: Normalized Significand before rounding

APPENDIX B
EXAMPLES OF SDDITION AND MULTIPLICATION

Let A= 24.25
B= -0.125

Then A is represented as S=0
M= 011000.01 = $1.100001 * 2^4$
E = 127 + 4 =131 where the bias is $2^{8-1} -1=127$

MSB(31)	30 ... 23	22 ... LSB(0)
S	EEEEEEEE	FFFFFFFFFFFFFFFFFFFFFFF

A=

MSB(31)	30 ... 23	22 ... LSB(0)
0	1000 0011	10000100000000000000000000

Then B is represented as S=1
M= 0.001 = $1.0000 * 2^{-3}$
E = 127 + (-3) =124 where the bias is $2^{8-1} -1 =127$

B=

MSB(31)	30 ... 23	22 ... LSB(0)
1	0111 1100	00000000000000000000000000

ADDITION
Now trying addition of these numbers

A+ B =R
Initially compare exponent of A to exponent of B and select the larger and note the difference.
eA > eB
and eA-eB =7
Now selecting the larger exponent to be the output exponent and shifting the smaller number by 7 bits to the right to align the binary point
Perform subtraction to obtain the significand of the reslt

$$\begin{array}{r} 1.1000010 * 2^4 \\ -\ 0.0000001 * 2^4 \\ \hline 1.10000001 * 2^4 \end{array}$$

$S_R = 0$
$e_R = 127 + 4 = 131$
$M_R = 1.10000001$

R=

MSB(31)	30 23	22 LSB(0)
0	1000 0011	10000010000000000000000000

Multiplication
Now trying multiplication of these numbers

A* B =R
Initially add the exponent of A to exponent of B and set the exponent of the results to this addition eA + eB
and eA+ eB = 1000 0011 + 0111 1100 – 0111 1111 = 1000 0000

Perform multiplication to obtain the significand of the result

$$\begin{array}{r} 1.1000010 \\ 0.0000001 \\ \hline 1.10000100 \end{array}$$

$S_R = 1$
$e_R = 1000\ 0000$
$M_R = 1.10000100$

R=

MSB(31)	30 … 23	22 … LSB(0)
1	1000 0000	10000100000000000000000