

Testing

Floating Point



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Fields					
COLII	CI I	Corner 2 Crop			
:	:	:	:	:	
:	:	:	:	•	
:	:	:	:	:	





Latitude/longitude



Fields					
Corner 1 Corner 2 Crop					
:	:	:	:	:	
:	:	:	:	:	
:	:	:	:	:	







Floating point numbers



Fields				
Corner 1 Corner 2 Crop				Crop
:	:	:	:	•
:	:	:	:	:
:	:	:	:	:







Latitude/longitude
Floating point numbers

That's when trouble starts

Fields				
Corner 1 Corner 2 Crop				Crop
:	:	:	:	•
:	:	:	:	:
:	:	:	:	:





Can't actually represent an infinite number of real values with a finite set of bit patterns



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What follows is (over-)simplified



Can't actually represent an infinite number of real values with a finite set of bit patterns
What follows is (over-)simplified

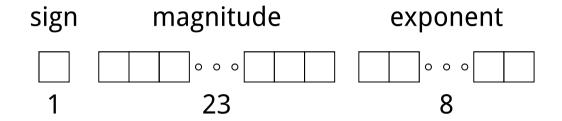
Goldberg (1991): "What Every Computer Scientist Should Know About Floating-Point Arithmetic"





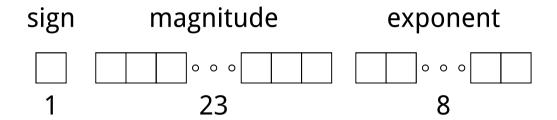
sign	magnitude	exponent
	0 0 0	000
1	23	8





To illustrate problems, we'll use a simpler format

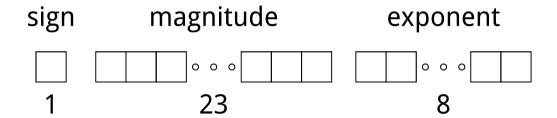




To illustrate problems, we'll use a simpler format

And only positive values without fractions





To illustrate problems, we'll use a simpler format And only positive values without fractions

magnitude	exponent
3	2



Exponent

Mantissa

	00	01	10	11
000	0	0	0	0
001	1	2	4	8
010	2	4	8	16
011	3	6	12	24
100	4	8	16	32
101	5	10	20	40
110	6	12	24	48
101	7	14	28	56



Exponent

Mantissa

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$$110 \times 2^{11}$$

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Mantissa

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$$\frac{110 \times 2^{11}}{6 \times 2^3}$$

Testing

Exponent

Mantissa

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$$\frac{110}{6 \times 2^3}$$

6 × 8

Testing

Exponent

Mantissa

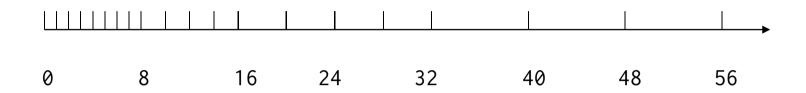
	00	01	10	11
000	0	0	0	0
001	1	2	4	8
010	2	4	8	16
011	3	6	12	24
100	4	8	16	32
101	5	10	20	40
110	6	12	24	48
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$$\frac{110}{6 \times 2^3}$$

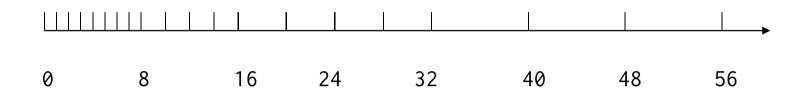
$$6 \times 2^{3}$$

Actual representation doesn't have redundancy



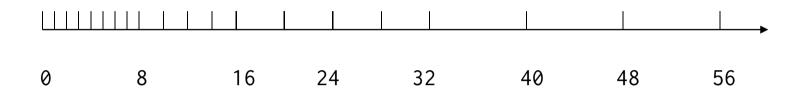






There are numbers we can't represent

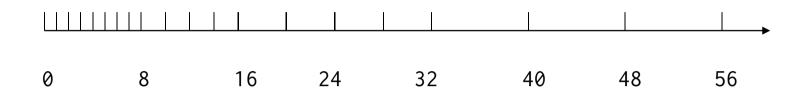




There are numbers we can't represent

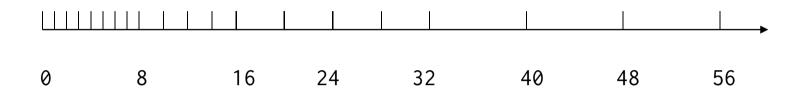
Just as 1/3 must be 0.3333 or 0.3334 in decimal





This scheme has no representation for 9

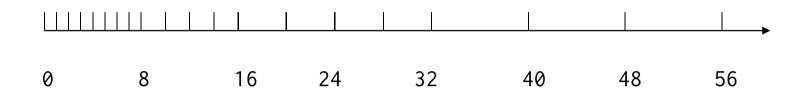




This scheme has no representation for 9

So 8+1 must be either 8 or 10



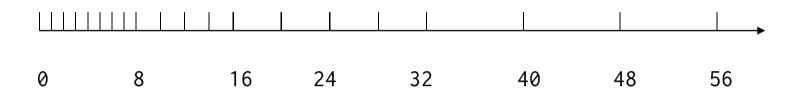


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So 8+1 must be either 8 or 10

If 8+1 = 8, what is 8+1+1?





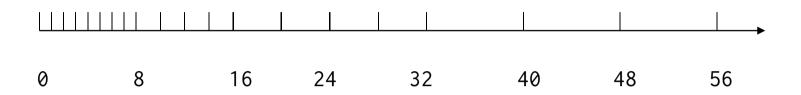
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(8+1)+1 = 8+1 (if we round down) = 8 again





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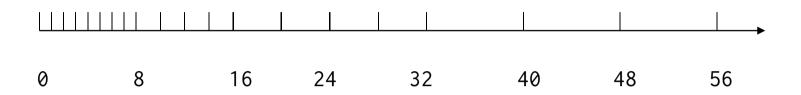
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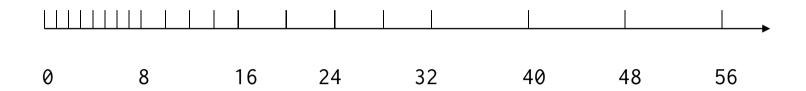
If 8+1 = 8, what is 8+1+1?

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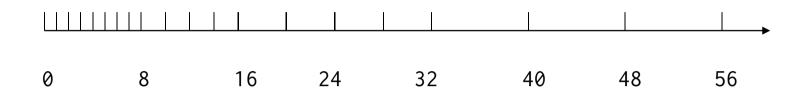
But 8+(1+1) = 8+2 = 10, which we *can* represent

"Sort then sum" would give the same answer...



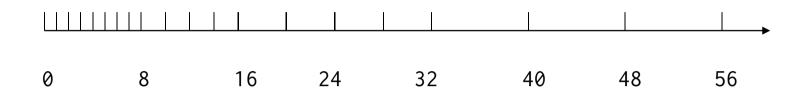






Spacing is uneven

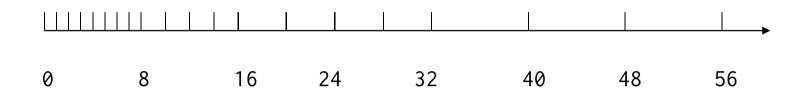




Spacing is uneven

But *relative* spacing stays the same





Spacing is uneven

But relative spacing stays the same

Because we're multiplying the same few mantissas by ever-larger exponents



Absolute error = |val - approx|



Absolute error = |val - approx|

Relative error = |val - approx| / val



Operation	Actual	Intended	Absolute	Relative
	Result	Result	Error	Error
8+1	8	9	1	1/9 (11.11%)
56+1	56	57	1	1/57 (1.75%)



Operation	Actual Result	Intended Result	Absolute Error	Relative Error
8+1	8	9	1	1/9 (11.11%)
56+1	56	57	1	1/57 (1.75%)

Relative error is more useful



Operation	Actual Result	Intended Result	Absolute Error	Relative Error
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56+1	56	57	1	1/57 (1.75%)

Relative error is more useful

Makes little sense to say "off by 0.01" if the value you're approximating is 0.00000001





```
vals = []
for i in range(1, 10):
  number = 9.0 * 10.0 ** -i
  vals.append(number)
  total = sum(vals)
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10): \leftarrow i = 1, 2, 3, ..., 9
  number = 9.0 * 10.0 ** -i
  vals.append(number)
  total = sum(vals)
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10):
                                      0.9, 0.09, 0.009, ...
  number = 9.0 * 10.0 ** −i ←
  vals.append(number)
  total = sum(vals)
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10):
  number = 9.0 * 10.0 ** -i
 vals.append(number)
  total = sum(vals) ←
                                      0.9, 0.99, 0.999, ...
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10):
  number = 9.0 * 10.0 ** -i
                                           But is it?
  vals.append(number)
  total = sum(vals) ←
                                      0.9, 0.99, 0.999, ...
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10):
  number = 9.0 * 10.0 ** -i
  vals.append(number)
                                        1-0.1, 1-0.01, ...
  total = sum(vals)
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
                                         Should also make
for i in range(1, 10):
  number = 9.0 \times 10.0 \times -i
                                         0.9, 0.99, ...
  vals.append(number)
                                         1-0.1, 1-0.01, ...
  total = sum(vals)
  expected = 1.0 - (1.0 * 10.0 ** i)
  diff = total - expected
  print '%2d %22.21f %22.21f' % \
        (i, total, total-expected)
```



```
vals = []
for i in range(1, 10):
  number = 9.0 * 10.0 ** -i
 vals.append(number)
  total = sum(vals)
 expected = 1.0 - (1.0 * 10.0 ** i)
 diff = total − expected ← Check and print
 print '%2d %22.21f %22.21f' % \
       (i, total, total-expected)
```



0.900000000000000022204 0.0000000000000000000000 0.9899999999999991118 0.0000000000000000000000 3 0.998999999999999112 0.0000000000000000000000 0.0000000000000000000000 0.999900000000000011013 4 5 0.0000000000000000000000 0.999990000000000045510 6 0.9999990000000000082267 0.000000000000000111022 0.99999990000000052636 0.000000000000000000000 8 0.999999990000000060775 0.000000000000000111022 9 0.999999999000000028282 0.0000000000000000000000



```
0.0000000000000000000000
  0.900000000000000022204
  0.989999999999991118
                      0.9989999999999999112
                      0.999900000000
                             200000000000000
  0.99999000000 Already slightly off
                             20000000000000
6
  0.999999000000000082267
                      0.000000000000000111022
  0.99999990000000052636
                      8
  0.999999990000000060775
                      0.000000000000000111022
9
                      0.999999999000000028282
```



```
0.900000000000000022204
                     0.0000000000000000000000
  0.989999999999991118
                     0.998999999999999112
                     0.999
                                   00000
        But at least they're consistent
  0.999
                                   00000
6
  0.999999000000000082267
                     0.000000000000000111022
  0.99999990000000052636
                     8
  0.999999990000000060775
                     0.000000000000000111022
9
  0.999999999000000028282
```



- 2 0.9899999999999991118 0.000000000000000000000
- 3 0.99899999999999999112 0.000000000000000000000
- 5 0.999990000000000045510 0.000000000000000000000
- 6 0.999999000000000082267 0.00000000000000111022
- 7 0.99999990000000052636 0.000000000000000000000

Sometimes, they're not



- 1 0.900000000000000022204
- 2 0.989999999999991118
- 3 0.998999999999999112
- 4 0.999900000000000011013
- 5 0.999990000000000045510
- 0.0000000000000000000000
- 6 0.999999000000000082267
- 0.000000000000000111022
- 7 0.99999990000000052636
- 8 0.99999999000000060775
- 0.00000000000000111022
- 9 0.99999999000000028282
- 0.00000000000000000000000



Sometimes errors cancel out later





What do you compare the actual test result to?



So what does this have to do with testing?
What do you compare the actual test result to?

sum(vals[0:7]) = 0.99999999 could well be False



What do you compare the actual test result to?

sum(vals[0:7]) = 0.9999999 could well be False

Even if vals = [0.9, 0.09, ...]



What do you compare the actual test result to?

sum(vals[0:7]) = 0.9999999 could well be False

Even if vals = [0.9, 0.09, ...]

And there are no bugs in our code



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sum(vals[0:7]) = 0.99999999 could well be False

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And there are no bugs in our code

This is *hard*



What do you compare the actual test result to?

sum(vals[0:7]) = 0.99999999 could well be False

Even if vals = [0.9, 0.09, ...]

And there are no bugs in our code

This is *hard*

No one has a good answer





1. Compare to analytic solutions (when you can)



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Use nose.assert_almost_equals(expected, actual)



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Use nose.assert_almost_equals(expected, actual)

Fail if the two objects are unequal as determined by their difference rounded to the given number of decimal places (default 7) and comparing to zero.



- 1. Compare to analytic solutions (when you can)
- 2. Compare complex to simple
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Use nose.assert_almost_equals(expected, actual)

Fail if the two objects are unequal as determined by their difference rounded to the given number of decimal places (default 7) and comparing to zero.

Is that absolute or relative?



created by

Greg Wilson

August 2010



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