Mysterious World of Python Float Arithmetics



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Agenda

- Problem Statement
- Float representations in base 10 and 2
- Float representation on hardware
- Python and Floats



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Guess the outcome



Expectations?

```
>>> 0.1 + 0.1 + 0.2 == 0.4
  TRUE
>>> 0.1 + 0.1 + 0.1 == 0.3
  TRUE
>>> 100 * 0.357
  35.7
>>> 64.8 - 35.8
  29.0
```

```
>>> 0.3 - 0.2 >= 0.1
 TRUF
>>> 500.50 - 400.05
 100.45
>>> 1000 * 0.357
 357
>>> 64.8 - 35.7
 29.1
```



Actual



What's Going On?

Fractions

- Fraction → Representing part/portion of whole thing
- $a/b \rightarrow a=Numerator$, b=Denominator
- ½, ¾, 9/10, 99/101



Decimals

- Type of numbers that have a whole and a fractional part, separated by a decimal point
- 3.1415, 9.98, 1.05, 34.901, etc.
- 3.1415
 - \circ 3 \rightarrow whole part
 - \circ 1415 \rightarrow fractional part
 - \circ . \rightarrow decimal point



Fractions and Decimals

- Both fractions and decimals are mostly interconvertible
 - o Focus here would be on Fraction -> Decimal
- Fraction -> Decimal
 - finite repeating decimals
 - infinite repeating decimals



Fractions and Decimals

- $\frac{1}{2} \rightarrow 0.50$
- $\frac{1}{4} \rightarrow 0.25$
- $\frac{1}{3} \rightarrow 0.333$ or 0.33333333 or 0.333333333
- % → 0.166666666667
- $3/11 \rightarrow 0.2727272727$



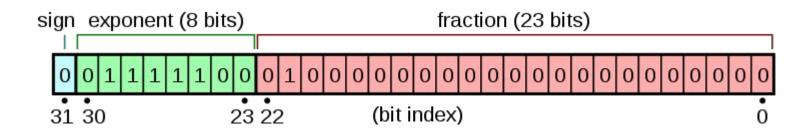
Float approximations - Base 2

- Like other numbers, floats represented as base-2 on hardware
- Finite and infinite repeating exists in binary as well
- $0.25 \rightarrow 0.01$
- $0.1 \rightarrow 0.00011001100110011001...$
- $0.1667 \rightarrow 0.00101010101011001101100111110100000...$



IEEE 754

- IEEE 754 standard used for 32 bits float representation
- Rounding off the fractional/mantissa to 23 bits leads to rounding off errors





- Representing 10.1 in IEEE 754
- Start with converting whole and fractional parts into binary
- 10
 - $0 10 / 2 \rightarrow \mathbf{R=0}, Q=5$
 - \circ 5/2 \to **R=1**, Q=2
 - \circ 2/2 \to **R=0**, Q=1
 - \circ 1/2 \to **R=1**, Q=0
- $10_{10} \rightarrow 1010_2$



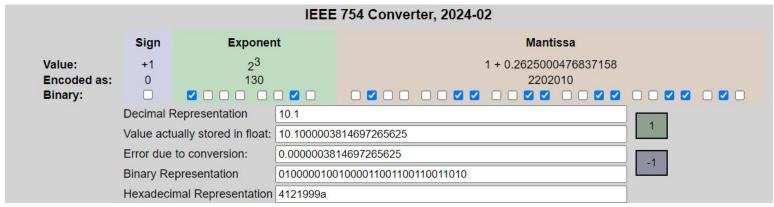
- 0.1
 - 0.1 * 2 = 0.2
 - 0.2 * 2 = 0.4
 - 0.4 * 2 = 0.8
 - o 0.8 * 2 = **1**.6
 - o 0.6 * 2 = **1**.2
 - 0.2 * 2 = 0.4
 - o 0.4 * 2 = **0**.8
 - o 0.8 * 2 = **1**.6
 - o 0.6 * 2 = **1**.2
 - o 0.2 * 2 = **0**.4
 - ... (infinite repeating)
- \bullet 0.1₁₀ = 000110011001100110011



- 1. $10.1_{10} = 1010.000110011001100110011$ (not done yet)
- 2. Next, shift the decimal either right or left until there is only '1' before decimal
 - a. $1010.0001100110011001100110011 \rightarrow 1.010000110011001100110011$
 - b. 3 points moved left
- 3. Round off decimal portion to 23 bits
 - a. **1.01000011001100110011001**10011 \rightarrow 1.010000110011001101010
- 4. Calculate biased exponent
 - a. IEEE 754, constant bias value = 127
 - b. Bias exponent calculated = $127 + 3 \rightarrow 130$ (add +3 as decimal point was shifted 3 points left, the value would be subtracted if shifted right)
 - c. $130_{10} = 10000010_2$



- 10.1
 - \circ Sign = 0
 - Exponent = 10000010
 - Mantissa = 0100001100110011001
- Assembled → 0 10000010 0100001100110011001





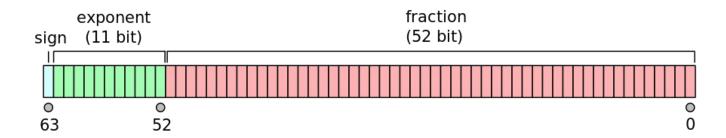
Summarized

- 1. Convert both whole and fractional parts to binary
- 2. Shift decimal so there is only one 1 in front of the point
- 3. Round off fractional part
- 4. Add bias to the exponent value
- 5. Convert exponent to the binary system
- 6. Assemble floating point number with sign bit at start



IEEE 754-2008 Standard

- Double precision floating point format
- Also known as float64 or binary64





- Floats represented as approximations in most cases
- A degree of precision error is there regardless
- Due to hardware, the approximation issue is part of nearly every programming language
- How do we work around that?



Workarounds in Python

- Fraction
- Decimal
- round
- isclose



Workarounds

- Built-in <u>fractions module</u>
- Represent rational numbers as Fractions
- perform arithmetics on fractions
- Format fractions to floats when needed



```
>>> fraction = Fraction(4, 5) + 2
                                       >>> fraction
>>> from fractions import Fraction
                                       Fraction(14, 5)
>>> fc = Fraction(2, 10)
                                       >>> f"{fraction:.2f}"
>>> fc + 1 == Fraction(6, 5)
                                        '2.80'
True
                                       >>> f"{fraction:.4f}"
                                        '2.8000'
                                     >>> Fraction('5/6')
                                     Fraction(5, 6)
     >>> Fraction(2) / 10
                                     >>> Fraction('9/50')
                                     Fraction(9, 50)
     Fraction(1, 5)
```

Fraction(9, 5)

Fraction(9, 5)

>>> Fraction(10, 100)

>>> Fraction(4, 5) + 1

Fraction(1, 10)

Fraction(9, 5)

>>> Fraction.from_float(1.80)

Fraction(8106479329266893, 4503599627370496)

>>> Fraction.from_float(1.80).limit_denominator(1000)

>>> Fraction.from_float(1.80).limit_denominator(100)

Workarounds

- Built-in <u>decimal module</u>
- Designed on same standards as floats but allow programmers to handle precisions explicitly
- Create module from various data types and perform arithmetics



```
Decimal('0.2857142857142857142857142857')
>>> getcontext().prec = 8
>>> Decimal(2)/7
Decimal('0.28571429')
>>>

>>> (Decimal(2)/7) * 4

Decimal('1.1428572')
>>> ((Decimal(2)/7) * 4).as_integer_ratio()
(2857143. 2500000)
```

Context(prec=28, rounding=ROUND_HALF_EVEN, Emin=-999999, Emax=999999, capitals=1, clamp=0, flags=[], traps=[InvalidOpera

>>> from decimal import *

tion, DivisionByZero, Overflow])

>>> getcontext()

>>> Decimal(2)/7

```
>>> getcontext()
Context(prec=8, rounding=ROUND_HALF_EVEN, Emin=-999999, Emax=999999, capitals=1, clamp=0, flags=[Inexact, FloatOperation, Rounded], traps=[InvalidOperation, DivisionByZero, Overflow])
>>> Decimal(3.1415)
Decimal('3.141500000000000181188397618825547397136688232421875')
>>> Decimal('3.1415904543565')
Decimal('3.1415904543565')
>>> Decimal(105)
Decimal('105)
Decimal('105')
>>> Decimal(2/7)
Decimal('0.28571428571428569842538536249776370823383331298828125')
>>> Decimal(2/7) + Decimal('3.1415904543565')
Decimal('3.4273047')
```

Workarounds

- Built-in <u>round function</u>
- Takes 2 arguments; number and precision
- Round off manually to avoid unexpected results



```
>>> round(0.1 + 0.1 + 0.1, 2) == round (0.3, 2)

True

>>> round(145.95-45.45, 1)

100.5
```

```
>>> round(0.1, 1) + round(0.1, 1) + round(0.1, 1) == round(0.3, 1)
False
>>> round(0.1 + 0.1 + 0.1, 1) == round(0.3, 1)
True
```

Workarounds

- Built-in <u>math.isclose</u>
- Verify if the provided values are close to each other
 - Closeness is calculated based on absolute and relative tolerance
- Relative tolerance → maximum allowed difference between values, greater than zero
- Absolute tolerance → minimum absolute difference
- Under the hood \rightarrow abs(a-b) <= max(rel_tol * max(abs(a), abs(b)), abs_tol).



```
>>> isclose(0.5, 0.3, rel_tol=0.05)
False
>>> isclose(0.5, 0.3, rel_tol=0.2)
False
>>> isclose(0.5, 0.3, rel_tol=0.4)
True
```

```
>>> isclose(0.6, 0.65, abs_tol=0.01, rel_tol=0.1)
True
>>> isclose(0.8, 0.65, abs_tol=0.01, rel_tol=0.1)
False
```

Unit Testing

- Unittest's built-in assertions
 - <u>assertAlmostEqual</u>
 - <u>assertNotAlmostEqual</u>
- Signature
 - First, second, decimal places (default 7), msg=None, delta=None
- How does it work?
 - o computing the difference of two numbers
 - rounding to the given number of decimal places
 - Comparing to zero
 - Raise assertion depending upon the type
- If delta is supplied, the difference between should be less or equal to or greater than delta.



```
# 0.12 - 0.09 <= 0.04
    self.assertAlmostEqual(0.12, 0.09, delta=0.04)
    \# \text{ round}(.501-0.50, 2) = 0
    self.assertAlmostEqual(0.501, 0.50, places=2)
    # round(0.50-0.4999, 3)
    self.assertAlmostEqual(0.4999, 0.50, places=3)
def test_almost_not_equal(self):
    \# round(0.501 - 0.50, 3) = 0.001
    self.assertNotAlmostEqual(0.501, 0.50, places=3)
    \# \text{ round}(0.50-0.4999, 4) = 0.0001
    self.assertNotAlmostEqual(0.4999, 0.50, places=4)
    # 0.01 >= 0.001
    self.assertNotAlmostEqual(0.5, 0.51, delta=0.001)
```

self.assertAlmostEqual(0.14, 0.15, delta=0.05)

def test_almost_equal(self):
 # 0.15-0.14 <= 0.05</pre>

Closing Thoughts

- Floating approximation is weird and can cause apps to behave unexpectedly
- Align or set expectations explicitly in the code using workarounds to avoid surprises



Helpful Links

- https://medium.com/python-in-plain-english/mysterious-world-of-pythons-floating-numbers-subtraction-42e157b4bd77
- https://nisal-pubudu.medium.com/how-to-deal-with-floating-point-rounding-err-or-5f77347a9549
- https://en.wikipedia.org/wiki/IEEE 754
- https://en.wikipedia.org/wiki/IEEE 754-2008 revision
- https://betterprogramming.pub/floating-point-numbers-are-weird-in-python-he-res-how-to-fix-them-51336e4ad51a
- https://docs.python.org/3/tutorial/floatingpoint.html
- https://jvns.ca/blog/2023/01/13/examples-of-floating-point-problems/
- https://numeral-systems.com/ieee-754-converter/
- https://www.h-schmidt.net/FloatConverter/IEEE754.html







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