# SimMethEx1

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# 1 Machine Epsilon

(a) float

```
#include <iostream>
int main(){
    float a = 1;
    float e = 1;
    float h = 1.1;
    while ((a + e) > a){
        std::cout<< (e) << std::endl;
        e = e/h;
}</pre>
```

### (b) double

```
double b = 1;
double eb = 1;
double hb = 1.1;
while ((b + eb) > b){
    std::cout<< (eb) << std::endl;
    eb = eb/hb;
}</pre>
```

data type	precision	$1.0 + \epsilon$
float double long double	5.7053e-08 1.05299e-16 5.14054e-20	1.000000063 1.000000000000000022 1.0000000000000000

Table 1: Results

#### (c) long double

```
long double c = 1;
long double ec = 1;
long double hc = 1.1;
while ((c + ec) > c){
    std::cout<< (ec) << std::endl;
    ec = ec/hc;
}
return 0;
}</pre>
```

## 2 Pitfalls of floating point arithmetic

```
double a = 1.0e17;
double b = -1.0e17;
double c = 1.0;
double x = (a + b) + c;
double y = a + (b + c);
cout << "x = " << x << ", " << "y = " << y << endl;
```

Result in the console: x = 1, y = 0

The precision for 1.0e17 is lower then for smaller numbers. In this case the precision is too small to take the 1.0 in to account.

## 3 Pitfalls of floating point representation

(a)

```
float xx = 0.01;
double yy = xx;
double zz = 0.01;

cout << "x = " << xx << ", " << "y = " << yy << ", z = " << zz << endl;
int i = xx * 10000;
int j = yy * 10000;
int k = zz * 10000;
printf("%d %d %d\n", i, j, k);</pre>
```

Result in the console: 100, 99, 1000

The dissenting value for (int j) results from the convert of the (float xx) into a (double yy). One might think a conversion from a lower to a higher precision number will not worsen the accuarcy of the value, however it turns out to be different.

- float  $a = 0.1 \rightarrow 0$  011111000 01000111101011100001010

By comparing b and c one notice that the better representation in b got additionally more ones in the mantissa.

(b)

float x = 0.01, so it has a 32 bit representation (1 bit sign S, 8 bit exponent E, 23 bit mantissa M).

x = 0 01111000 01000111101011100001010  $\rightarrow S = 0$ , E = 120, M = 2348810

$$x = (-1)^S 2^{E-b} \left( 1 + \frac{M}{2^p} \right) \tag{1}$$

$$=2^{-7}\left(1+\frac{2348810}{2^{23}}\right) \tag{2}$$

$$=\frac{2^{23} + 2348810}{2^{30}} = \frac{10737418}{1073741824} = \frac{5368709}{536870912}$$
(3)

where p = 23 and b = 127

## 4 Packing of numbers

#### Lösungsvorschlag 1

#### Lösungsvorschlag 2

```
int main(){
    float x = 1;
    float e = 1;
    float h = 1.1;
    while ((x + e) > x){
        e = e/h;
    }
    std::cout<< (e) << std::endl;

float y = 255;
    float e = 1;
    float h = 1.1;
    while ((y + e) > y){
        e = e/h;
    }
    std::cout<< (e) << std::endl;</pre>
```

This code returns the minimal resolution for float  $\mathbf{x}=1$  and float  $\mathbf{y}=255,$  which is

- $res(x) \sim 10^{-8}$
- $res(y) \sim 10^{-6}$

Therefor the intervall between 1 &2 contains  $\sim 10^8$  and the intevall between 255 & 256  $\sim 10^6$  numbers.

For double precision the code is equivalent and returns:

- $res(x) \sim 10^{-16}$
- $res(y) \sim 10^{-14}$

Therefor the intervall between 1 &2 contains  $\sim 10^16$  and the intevall between 255 & 256  $\sim 10^14$  numbers.

Hier https://www.h-schmidt.net/FloatConverter/IEEE754.html ist so ein IEEE 754 converter. Dort wird die Zahl als flaot auch immer nur auf ca die oben angegebene Genauigkeit angegeben und mit entsprechend mehr Kommastellen als double, daher denke ich das Programm rechnet richtig...

### 5 Summing a long list of numbers

Results from Python:

Sum beginning to end: -1.27636535808e+89 Sum end to beginning: 1.60798539786e-190 Sorted sum most neg to most pos: nan Sorted sum (absolute values): nan

was machen wir bei der 4?, Erklärung bei der 3 a!

### Exercise\_1.5

#### October 26, 2016

```
In [1]: import numpy as np
In [2]: f = open('numbers.dat', 'rb') # opens numbers file
        data = np.fromfile(f, dtype = float) # read binary data out of file
        data128 = np.fromfile(f, dtype = np.float128) # now with long double data type
In [3]: len(data) #make sure it's a million numbers
Out[3]: 1000000
In [4]: ### Part a, sum from beginning to end ###
       sumup = 0.
       for i in range(len(data)):
            sumup =+ data[i]
       print('Sum (a) = ', sumup)
Sum (a) = -1.27636535808e+89
In [5]: ### Part b, reversed direction of summing ###
        sumdown = 0.
        for i in reversed(range(len(data))):
            sumdown =+ data[i]
       print('Sum (b) = ', sumdown)
Sum (b) = 1.60798539786e-190
In [6]: ### Part c, sort by magnitude (does that mean most negative or the smallest absolute value?) an
        # Sort data
       datasort = np.sort(data) # sort from most negative to most positive
       datasortabs = np.sort(np.abs(data)) # sort from smallest absolute to highest absolute
        # Check data types
       print('Datasort type: ',datasort.dtype, '\nDatasort absolute data type: ', datasortabs.dtype)
        # Sum
        sumsortabs = 0.
       sumsort = 0.
       for i in range(len(datasort)):
            sumsort =+ datasort[i]
            sumsortabs =+ datasortabs[i]
       print('Sum (c) =', sumsort, '\nSum (c) absolute =', sumsortabs)
Datasort type: float64
Datasort absolute data type: float64
Sum (c) = nan
Sum (c) absolute = nan
```

```
In [7]: print(np.sum(np.isnan(data)))
526
/home/sophia/anaconda3/lib/python3.4/site-packages/IPython/kernel/_main_.py:1: RuntimeWarning: invalid
  if __name__ == '__main__':
  There are 526 nan values. In both sorted arrays the nan values are at the end of the array. Operations
with these values will again give nan results. Therefore both sorted sums are nan. The nan values could be
below the double precision.
In [8]: print(np.min(data[np.isnan(data) ==False]), np.max(data[np.isnan(data) == False]))
-1.78701556856e+308 1.79005606756e+308
/home/sophia/anaconda3/lib/python3.4/site-packages/IPython/kernel/_main_.py:1: RuntimeWarning: invalid
  if __name__ == '__main__':
  The smallest and biggest numbers which are not nan are near to the smallest representable numbers in
double precision. A higher precision could avoid the nan values.
In [9]: ### Part d, as c only with long double data ###
        # Sort data
        datasort128 = np.sort(data128) # sort from most negative to most positive
        datasortabs128 = np.sort(np.abs(data128)) # sort from smallest absolute to highest absolute
        # Check data types
        print('Datasort type: ',datasort128.dtype, '\nDatasort absolute data type: ', datasortabs128.dt
        # Sum
        sumsortabs128 = 0.
        sumsort128 = 0.
        for i in range(len(datasort128)):
            sumsort128 =+ datasort128[i]
            sumsortabs128 =+ datasortabs128[i]
        print('Sum (d) =', sumsort128, '\nSum (d) absolute =', sumsortabs128)
Datasort type: float128
Datasort absolute data type: float128
Sum (d) = 0.0
Sum (d) absolute = 0.0
In [10]: print('Sum beginning to end: ', sumup, '\nSum end to beginning: ', \
               sumdown, '\nSorted sum most neg to most pos: ',sumsort, '\nSorted sum (absolute values):
              '\nSorted sum long double: ', sumsort128, '\nSorted sum long double (abs values): ', sums
Sum beginning to end: -1.27636535808e+89
Sum end to beginning: 1.60798539786e-190
Sorted sum most neg to most pos: nan
Sorted sum (absolute values): nan
Sorted sum long double: 0.0
```

Sorted sum long double (abs values): 0.0