#### App. A: Sequences and difference equations

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#### Sequences

Sequences is a central topic in mathematics:

$$x_0, x_1, x_2, \ldots, x_n, \ldots,$$

Example: all odd numbers

$$1, 3, 5, 7, \ldots, 2n + 1, \ldots$$

For this sequence we have a formula for the *n*-th term:

$$x_n = 2n + 1$$

and we can write the sequence more compactly as

$$(x_n)_{n=0}^{\infty}, \quad x_n = 2n+1$$

## Other examples of sequences

1, 4, 9, 16, 25, ... 
$$(x_n)_{n=0}^{\infty}$$
,  $x_n = n^2$   
1,  $\frac{1}{2}$ ,  $\frac{1}{3}$ ,  $\frac{1}{4}$ , ...  $(x_n)_{n=0}^{\infty}$ ,  $x_n = \frac{1}{n+1}$   
1, 1, 2, 6, 24, ...  $(x_n)_{n=0}^{\infty}$ ,  $x_n = n!$ 

$$1, 1+x, 1+x+\frac{1}{2}x^2, 1+x+\frac{1}{2}x^2+\frac{1}{6}x^3, \dots (x_n)_{n=0}^{\infty}, x_n=\sum_{j=0}^n \frac{x^j}{j!}$$

## Finite and infinite sequences

- Infinite sequences have an infinite number of terms  $(n \to \infty)$
- In mathematics, infinite sequences are widely used
- In real-life applications, sequences are usually finite:  $(x_n)_{n=0}^N$
- Example: number of approved exercises every week in INF1100  $x_0, x_1, x_2, \dots, x_{15}$
- Example: the annual value of a loan  $x_0, x_1, \ldots, x_{20}$

## Difference equations

- For sequences occurring in modeling of real-world phenomena, there is seldom a formula for the *n*-th term
- However, we can often set up one or more equations governing the sequence
- Such equations are called difference equations
- With a computer it is then very easy to generate the sequence by solving the difference equations
- Difference equations have lots of applications and are very easy to solve on a computer, but often complicated or impossible to solve for x<sub>n</sub> (as a formula) by pen and paper!
- The programs require only loops and arrays

## Modeling interest rates

#### Problem:

Put  $x_0$  money in a bank at year 0. What is the value after N years if the interest rate is p percent per year?

#### Solution:

The fundamental information relates the value at year n,  $x_n$ , to the value of the previous year,  $x_{n-1}$ :

$$x_n = x_{n-1} + \frac{p}{100} x_{n-1}$$

How to solve for  $x_n$ ? Start with  $x_0$ , compute  $x_1, x_2, ...$ 

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## Simulating the difference equation for interest rates

#### What does it mean to simulate?

Solve math equations by repeating a simple procedure (relation) many times (boring, but well suited for a computer!)

```
Program for x_n = x_{n-1} + (p/100)x_{n-1}:
    from scitools.std import *
    x0 = 100
                                    # initial amount
    p = 5
                                    # interest rate
                                    # number of years
    index_set = range(N+1)
    x = zeros(len(index set))
    # Solution:
    x[0] = x0
    for n in index_set[1:]:
        x[n] = x[n-1] + (p/100.0)*x[n-1]
    print x
    plot(index_set, x, 'ro', xlabel='years', ylabel='amount')
```

# We do not need to store the entire sequence, but it is convenient for programming and later plotting

- Previous program stores all the  $x_n$  values in a NumPy array
- To compute  $x_n$ , we only need one previous value,  $x_{n-1}$

Thus, we could only store the two last values in memory:

```
x_old = x0
for n in index_set[1:]:
    x_new = x_old + (p/100.)*x_old
    x_old = x_new # x_new becomes x_old at next step
```

However, programming with an array x[n] is simpler, safer, and enables plotting the sequence, so we will continue to use arrays in the examples

## Daily interest rate

- A more relevant model is to add the interest every day
- The interest rate per day is r = p/D if p is the annual interest rate and D is the number of days in a year
- A common model in business applies D=360, but n counts exact (all) days

Just a minor change in the model:

$$x_n = x_{n-1} + \frac{r}{100} x_{n-1}$$

How can we find the number of days between two dates?

```
>>> import datetime

>>> date1 = datetime.date(2007, 8, 3)  # Aug 3, 2007

>>> date2 = datetime.date(2008, 8, 4)  # Aug 4, 2008

>>> diff = date2 - date1

>>> print diff.days

367
```

## Program for daily interest rate

```
from scitools.std import *
                                    # initial amount
x0 = 100
p = 5
                                    # annual interest rate
r = p/360.0
                                    # daily interest rate
import datetime
date1 = datetime.date(2007, 8, 3)
date2 = datetime.date(2011, 8, 3)
diff = date2 - date1
N = diff.days
index_set = range(N+1)
x = zeros(len(index_set))
# Solution:
x[0] = x0
for n in index_set[1:]:
    x[n] = x[n-1] + (r/100.0)*x[n-1]
print x
plot(index_set, x, 'ro', xlabel='days', ylabel='amount')
```

## But the annual interest rate may change quite often...

#### Varying p means $p_n$ :

- Could not be handled in school (cannot apply  $x_n = x_0 (1 + \frac{p}{100})^n$ )
- A varying p causes no problems in the program: just fill an array p with correct interest rate for day n

#### Modified program:

# Payback of a loan

- A loan L is paid back with a fixed amount L/N every month over N months + the interest rate of the loan
- ullet p: annual interest rate, p/12 : monthly rate
- Let  $x_n$  be the value of the loan at the end of month n

The fundamental relation from one month to the text:

$$x_n = x_{n-1} + \frac{p}{12 \cdot 100} x_{n-1} - (\frac{p}{12 \cdot 100} x_{n-1} + \frac{L}{N})$$

which simplifies to

$$x_n = x_{n-1} - \frac{L}{N}$$

(L/N makes the equation nonhomogeneous)

# How to make a living from a fortune with constant consumption

- We have a fortune F invested with an annual interest rate of p percent
- Every year we plan to consume an amount  $c_n$  (n counts years)
- Let  $x_n$  be our fortune at year n

A fundamental relation from one year to the other is

$$x_n = x_{n-1} + \frac{p}{100}x_{n-1} - c_n$$

Simplest possibility: keep  $c_n$  constant, but inflation demands  $c_n$  to increase...

# How to make a living from a fortune with inflation-adjusted consumption

- Assume I percent inflation per year
- Start with  $c_0$  as q percent of the interest the first year
- c<sub>n</sub> then develops as money with interest rate I

 $x_n$  develops with rate p but with a loss  $c_n$  every year:

$$x_n = x_{n-1} + \frac{p}{100}x_{n-1} - c_{n-1}, \quad x_0 = F, \ c_0 = \frac{pq}{10^4}F$$

$$c_n = c_{n-1} + \frac{I}{100}c_{n-1}$$

This is a coupled system of *two* difference equations, but the programming is still simple: we update two arrays, first x[n], then c[n], inside the loop (good exercise!)

#### The mathematics of Fibonacci numbers

No programming or math course is complete without an example on Fibonacci numbers:

$$x_n = x_{n-1} + x_{n-2}, \quad x_0 = 1, \ x_1 = 1$$

#### Mathematical classification

This is a homogeneous difference equation of second order (second order means three levels: n, n-1, n-2). This classification is important for mathematical solution technique, but not for simulation in a program.

Fibonacci derived the sequence by modeling rat populations, but the sequence of numbers has a range of peculiar mathematical properties and has therefore attracted much attention from mathematicians.

# Program for generating Fibonacci numbers

```
N = int(sys.argv[1])
from numpy import zeros
x = zeros(N+1, int)
x[0] = 1
x[1] = 1
for n in range(2, N+1):
    x[n] = x[n-1] + x[n-2]
    print n, x[n]
```

## Fibonacci numbers can cause overflow in NumPy arrays

#### Run the program with N = 50:

```
2 2
3 3
4 5
5 8
6 13
...
45 1836311903
Warning: overflow encountered in long_scalars
46 -1323752223
```

#### Note:

- Changing int to long or int 64 for array elements allows  $N \leq 91$
- Can use float96 (though  $x_n$  is integer):  $N \le 23600$

## No overflow when using Python int types

- Best: use Python scalars of type int these automatically changes to long when overflow in int
- The long type in Python has arbitrarily many digits (as many as required in a computation!)
- Note: long for arrays is 64-bit integer (int64), while scalar long in Python is an integer with as "infinitely" many digits

# Program with Python's int type for integers

The program now avoids arrays and makes use of three int objects (which automatically changes to long when needed):

```
import sys
    N = int(sys.argv[1])
    xnm1 = 1
                                         # "x_n minus 1"
    xnm2 = 1
                                         # "x_n minus 2"
    n = 2
    while n \le N:
        xn = xnm1 + xnm2
        print x_{d} = d' \% (n, xn)
        xnm2 = xnm1
        xnm1 = xn
        n += 1
Run with N = 200:
    x 2 = 2
    x 3 = 3
    x 198 = 173402521172797813159685037284371942044301
    x 199 = 280571172992510140037611932413038677189525
    \mathbf{x} 200 = 453973694165307953197296969697410619233826
```

Limition: your computer's memory

# New problem setting: exponential growth with limited environmental resources

The model for growth of money in a bank has a solution of the type

$$x_n = x_0 C^n \quad (= x_0 e^{n \ln C})$$

#### Note:

- This is exponential growth in time (n)
- Populations of humans, animals, and cells also exhibit the same type of growth as long as there are unlimited resources (space and food)
- Most environments can only support a maximum number M of individuals
- How can we model this limitation?

# Modeling growth in an environment with limited resources

Initially, when there are enough resources, the growth is exponential:

$$x_n = x_{n-1} + \frac{r}{100} x_{n-1}$$

The growth rate r must decay to zero as  $x_n$  approaches M. The simplest variation of r(n) is a linear:

$$r(n) = \varrho \left( 1 - \frac{x_n}{M} \right)$$

Observe:  $r(n) \approx \varrho$  for small n when  $x_n \ll M$ , and  $r(n) \to 0$  as  $x_n \to M$  and n is big

#### Logistic growth model:

$$x_n = x_{n-1} + \frac{\varrho}{100} x_{n-1} \left( 1 - \frac{x_{n-1}}{M} \right)$$

(This is a nonlinear difference equation)

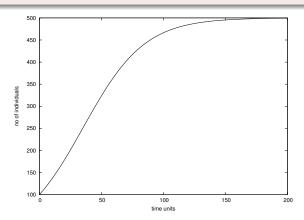
#### The evolution of logistic growth

In a program it is easy to introduce logistic instead of exponential growth, just replace

$$x[n] = x[n-1] + p/100.0)*x[n-1]$$

by

$$x[n] = x[n-1] + (rho/100.0)*x[n-1]*(1 - x[n-1]/float(M))$$



## The factorial as a difference equation

The factorial n! is defined as

$$n(n-1)(n-2)\cdots 1, \quad 0!=1$$

The following difference equation has  $x_n = n!$  as solution and can be used to compute the factorial:

$$x_n = nx_{n-1}, \quad x_0 = 1$$

## Difference equations must have an initial condition

- In mathematics, it is much stressed that a difference equation for  $x_0$  must have an initial condition  $x_0$
- The initial condition is obvious when programming: otherwise we cannot start the program  $(x_0 \text{ is needed to compute } x_n)$
- However: if you forget x[0] = x0 in the program, you get  $x_0 = 0$  (because x = zeroes(N+1)), which (usually) gives unintended results!

# Have you ever though about how $\sin x$ is really calculated?

- How can you calculate sin x, ln x, ex without a calculator or program?
- These functions were originally defined to have some desired mathematical properties, but without an algorithm for how to evaluate function values
- Idea: approximate sin x, etc. by polynomials, since they are easy to calculate (sum, multiplication), but how??

## Would you expect these fantastic mathematical results?

#### Amazing result by Gregory, 1667:

$$\sin x = \sum_{k=0}^{\infty} (-1)^k \frac{x^{2k+1}}{(2k+1)!}$$

#### Even more amazing result by Taylor, 1715:

$$f(x) = \sum_{k=0}^{\infty} \frac{1}{k!} \left( \frac{d^k}{dx^k} f(0) \right) x^k$$

For "any" f(x), if we can differentiate, add, and multiply  $x^k$ , we can evaluate f at any x (!!!)

#### Taylor polynomials

#### Practical applications works with a truncated sum:

$$f(x) \approx \sum_{k=0}^{N} \frac{1}{k!} \left( \frac{d^{k}}{dx^{k}} f(0) \right) x^{k}$$

 ${\it N}=1$  is  ${\it very}$  popular and has been essential in developing physics and technology

#### Example:

$$e^{x} = \sum_{k=0}^{\infty} \frac{x^{k}}{k!}$$

$$\approx 1 + x + \frac{1}{2}x^{2} + \frac{1}{6}x^{3}$$

$$\approx 1 + x$$

# Taylor polynomials around an arbitrary point

The previous Taylor polynomials are most accurate around x = 0. Can make the polynomials accurate around any point x = a:

$$f(x) \approx \sum_{k=0}^{N} \frac{1}{k!} \left( \frac{d^k}{dx^k} f(a) \right) (x - a)^k$$

## Taylor polynomial as one difference equation

The Taylor series for  $e^x$  around x = 0 reads

$$e^{x} = \sum_{n=0}^{\infty} \frac{x^{n}}{n!}$$

Define

$$e_n = \sum_{k=0}^{n-1} \frac{x^k}{k!} = \sum_{k=0}^{n-2} \frac{x^k}{k!} + \frac{x^{n-1}}{(n-1)!}$$

We can formulate the sum in  $e_n$  as the following difference equation:

$$e_n = e_{n-1} + \frac{x^{n-1}}{(n-1)!}, \quad e_0 = 0$$

# More efficient computation: the Taylor polynomial as two difference equations

Observe:

$$\frac{x^n}{n!} = \frac{x^{n-1}}{(n-1)!} \cdot \frac{x}{n}$$

Let  $a_n = x^n/n!$ . Then we can efficiently compute  $a_n$  via

$$a_n = a_{n-1} \frac{x}{n}, \quad a_0 = 1$$

Now we can update each term via the  $a_n$  equation and sum the terms via the  $e_n$  equation:

$$e_n = e_{n-1} + a_{n-1}, \quad e_0 = 0, \ a_0 = 1$$
  
 $a_n = \frac{x}{n} a_{n-1}$ 

See the book for more details

# Nonlinear algebraic equations

#### Generic form of any (algebraic) equation in x:

$$f(x) = 0$$

#### Examples that can be solved by hand:

$$ax + b = 0$$
$$ax^{2} + bx + c = 0$$
$$\sin x + \cos x = 1$$

- Simple numerical algorithms can solve "any" equation f(x) = 0
- Safest: Bisection
- Fastest: Newton's method
- Don't like f'(x) in Newton's method? Use the Secant method
- Secant and Newton are difference equations!

## Newton's method for finding zeros

#### Newton's method

Simpson (1740) came up with the following general method for solving f(x) = 0 (based on ideas by Newton):

$$x_n = x_{n-1} - \frac{f(x_{n-1})}{f'(x_{n-1})}, \quad x_0 \text{ given}$$

#### Note:

- This is a (nonlinear!) difference equation
- As  $n \to \infty$ , we hope that  $x_n \to x_s$ , where  $x_s$  solves  $f(x_s) = 0$
- How to choose N when what we want is  $x_N$  close to  $x_s$ ?
- Need a slightly different program: simulate until  $f(x) \le \epsilon$ , where  $\epsilon$  is a small tolerance
- ullet Caution: Newton's method may (easily) diverge, so  $f(x) \leq \epsilon$  may never occur!

## A program for Newton's method

#### Quick implementation:

```
def Newton(f, x, dfdx, epsilon=1.0E-7, max_n=100):
    n = 0
    while abs(f(x)) > epsilon and n <= max_n:
        x = x - f(x)/dfdx(x)
        n += 1
    return x, n, f(x)</pre>
```

#### Note:

- f(x) is evaluated twice in each pass of the loop only one evaluation is strictly necessary (can store the value in a variable and reuse it)
- f(x)/dfdx(x) can give integer division
- It could be handy to store the x and f(x) values in each iteration (for plotting or printing a convergence table)

## An improved function for Newton's method

```
Only one f(x) call in each iteration, optional storage of (x, f(x))
values during the iterations, and ensured float division:
    def Newton(f, x, dfdx, epsilon=1.0E-7, max_n=100,
                store=False):
        f_value = f(x)
        \mathbf{n} = 0
        if store: info = [(x, f_value)]
        while abs(f_value) > epsilon and n <= max_n:
             x = x - float(f_value)/dfdx(x)
             n += 1
             f_value = f(x)
             if store: info.append((x, f_value))
        if store:
             return x, info
        else:
             return x, n, f_value
```

## Application of Newton's method

$$e^{-0.1x^2}\sin(\frac{\pi}{2}x)=0$$

Solutions:  $x = 0, \pm 2, \pm 4, \pm 6, ...$ 

```
Main program:
    from math import sin, cos, exp, pi
    import sys
    def g(x):
        return \exp(-0.1*x**2)*\sin(pi/2*x)
    def dg(x):
        return -2*0.1*x*exp(-0.1*x**2)*sin(pi/2*x) + 
               pi/2*exp(-0.1*x**2)*cos(pi/2*x)
    x0 = float(sys.argv[1])
    x, info = Newton(g, x0, dg, store=True)
    print 'Computed zero:', x
    # Print the evolution of the difference equation
    # (i.e., the search for the root)
    for i in range(len(info)):
        print 'Iteration %3d: f(%g)=%g' % (i, info[i][0], info[i][1])
```

## Results from this test problem

```
x_0=1.7 gives quick convergence towards the closest root x=0:

zero: 1.999999999768449

Iteration 0: f(1.7)=0.340044

Iteration 1: f(1.99215)=0.00828786

Iteration 2: f(1.99998)=2.53347e-05

Iteration 3: f(2)=2.43808e-10

Start value x_0=3 (closest root x=2 or x=4):

zero: 42.49723316011362

Iteration 0: f(3)=-0.40657

Iteration 1: f(4.66667)=0.0981146

Iteration 2: f(42.4972)=-2.59037e-79
```

## What happened here??

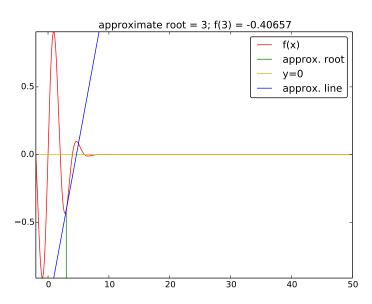
Try the demo program  $src/diffeq/Newton\_movie.py$  with  $x_0 = 3$ ,  $x \in [-2, 50]$  for plotting and numerical approximation of f'(x):

Terminal> python Newton\_movie.py "exp(-0.1\*x\*\*2)\*sin(pi/2\*x)" numeric 3 -2 50

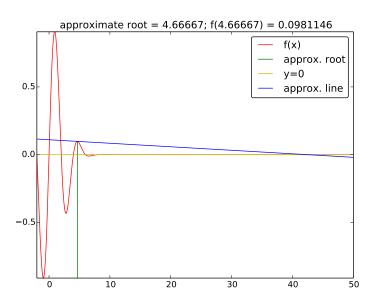
#### Lesson learned:

Newton's method may work fine or give wrong results! You need to understand the method to interpret the results!

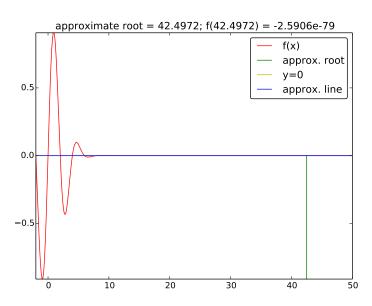
## First step: we're moving to the right (x = 4?)



## Second step: oops, too much to the right...



# Third step: disaster since we're "done" $(f(x) \approx 0)$



# Programming with sound

#### Tones are sine waves:

A tone A (440 Hz) is a sine wave with frequency 440 Hz:

$$s(t) = A \sin(2\pi f t), \quad f = 440$$

On a computer we represent s(t) by a discrete set of points on the function curve (exactly as we do when we plot s(t)). CD quality needs 44100 samples per second.

# Making a sound file with single tone (part 1)

- r: sampling rate (samples per second, default 44100)
- f: frequency of the tone
- m: duration of the tone (seconds)

Sampled sine function for this tone:

$$s_n = A \sin \left(2\pi f \frac{n}{r}\right), \quad n = 0, 1, \dots, m \cdot r$$

Code (we use descriptive names: frequency f, length m, amplitude A, sample\_rate r):

# Making a sound file with single tone (part 2)

- We have data as an array with float and unit amplitude
- Sound data in a file should have 2-byte integers (int16) as data elements and amplitudes up to  $2^{15}-1$  (max value for int16 data)

```
data = note(440, 2)
data = data.astype(numpy.int16)
max_amplitude = 2**15 - 1
data = max_amplitude*data
import scitools.sound
scitools.sound.write(data, 'Atone.wav')
scitools.sound.play('Atone.wav')
```

## Reading sound from file

- Let us read a sound file and add echo
- Sound = array s[n]
- Echo means to add a delay of the sound

### Load data, add echo and play:

```
data = scitools.sound.read(filename)
data = data.astype(float)
data = add_echo(data, beta=0.6)
data = data.astype(int16)
scitools.sound.play(data)
```

## Playing many notes

- Each note is an array of samples from a sine with a frequency corresponding to the note
- Assume we have several note arrays data1, data2, ...:

```
# put data1, data2, ... after each other in a new array:
data = numpy.concatenate((data1, data2, data3, ...))

The start of "Nothing Else Matters" (Metallica):

E1 = note(164.81, .5)
G = note(392, .5)
B = note(493.88, .5)
E2 = note(659.26, .5)
intro = numpy.concatenate((E1, G, B, E2, B, G))
...
song = numpy.concatenate((intro, intro, ...))
scitools.sound.play(song)
scitools.sound.write(song, 'tmp.wav')
```

## Summary of difference equations

- Sequence:  $x_0, x_1, x_2, ..., x_n, ..., x_N$
- Difference equation: relation between  $x_n$ ,  $x_{n-1}$  and maybe  $x_{n-2}$  (or more terms in the "past") + known start value  $x_0$  (and more values  $x_1$ , ... if more levels enter the equation)

#### Solution of difference equations by simulation:

```
index_set = <array of n-values: 0, 1, ..., N>
x = zeros(N+1)
x[0] = x0
for n in index_set[1:]:
    x[n] = <formula involving x[n-1]>
```

Can have (simple) systems of difference equations:

```
for n in index_set[1:]:
    x[n] = <formula involving x[n-1]>
    y[n] = <formula involving y[n-1] and x[n]>
```

Taylor series and numerical methods such as Newton's method can be formulated as difference equations, often resulting in a good way of programming the formulas

## Summarizing example: music of sequences

- Given a  $x_0, x_1, x_2, ..., x_n, ..., x_N$
- Can we listen to this sequence as "music"?
- Yes, we just transform the  $x_n$  values to suitable frequencies and use the functions in scitools.sound to generate tones

We will study two sequences:

$$x_n = e^{-4n/N} \sin(8\pi n/N)$$

and

$$x_n = x_{n-1} + qx_{n-1} (1 - x_{n-1}), \quad x = x_0$$

The first has values in [-1,1], the other from  $x_0=0.01\ \mathrm{up}$  to around 1

Transformation from "unit"  $x_n$  to frequencies:

$$y_n = 440 + 200x_n$$

(first sequence then gives tones between 240 Hz and 640 Hz)

## Module file: soundeq.py

- Three functions: two for generating sequences, one for the sound
- Look at files/soundeq.py for complete code

#### Try it out in these examples:

```
Terminal> python soundseq.py oscillations 40 Terminal> python soundseq.py logistic 100
```

Try to change the frequency range from 200 to 400.