

# Genome 540 Discussion

Conor Camplisson

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# Outline

- Homework 1 wrap-up
- C/C++/Python programming tips
- Homework 2 overview

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# Homework #1 Wrap-up

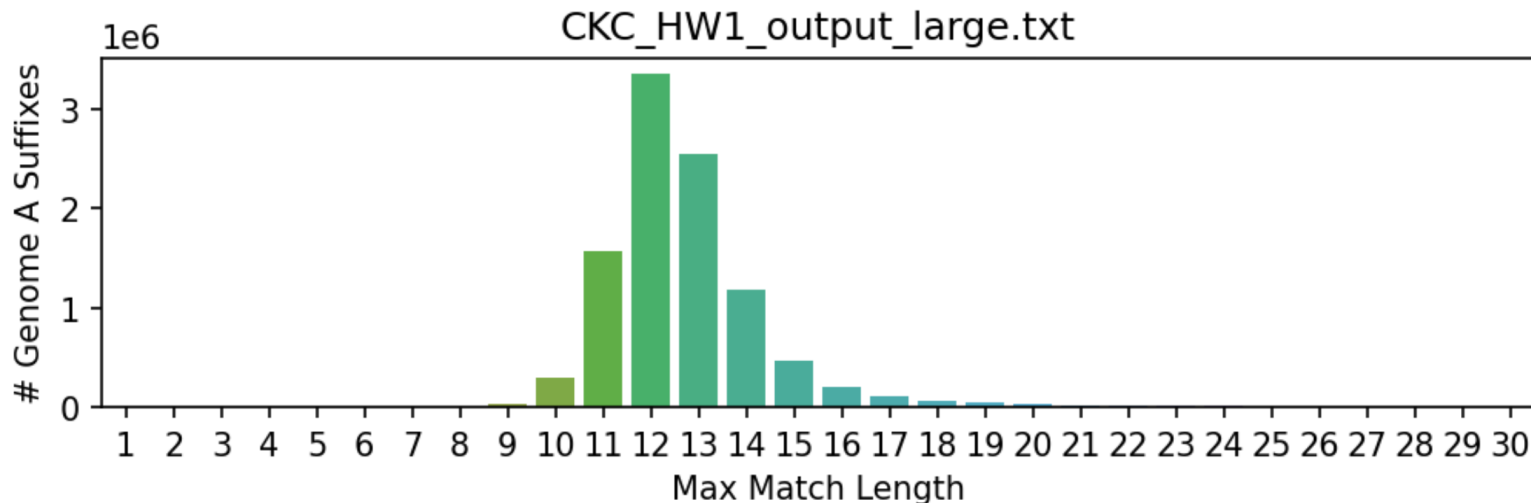
p<sub>10</sub> AAACCGTACACTGGGTTCAAGAGATTTCCC  
p<sub>11</sub> AACCGTACACTGGGTTCAAGAGATTTCCC  
p<sub>28</sub> AAGAGATTTCCC  
p<sub>17</sub> ACACTGGGTTCAAGAGATTTCCC  
p<sub>12</sub> ACCGTACACTGGGTTCAAGAGATTTCCC  
p<sub>1</sub> ACCTGCACTAAACCGTACACTGGGTTCAAGAGATTTCCC  
p<sub>7</sub> ACTAAACCGTACACTGGGTTCAAGAGATTTCCC  
p<sub>19</sub> ACTGGGTTCAAGAGATTTCCC  
p<sub>29</sub> AGAGATTTCCC  
p<sub>31</sub> AGATTTCCC  
p<sub>33</sub> ATTTCCC  
p<sub>27</sub> CAAGAGATTTCCC  
⋮

## Common bugs

- Hist values slightly off
  - Sorting logic, sorted array processing logic
- Genomic coordinate 1-index

## Common debug scenarios

- Too much memory usage
  - Storing substrings (silent caching in python)
- Sort step is too slow
  - Slow comparison function
  - Many comparisons!

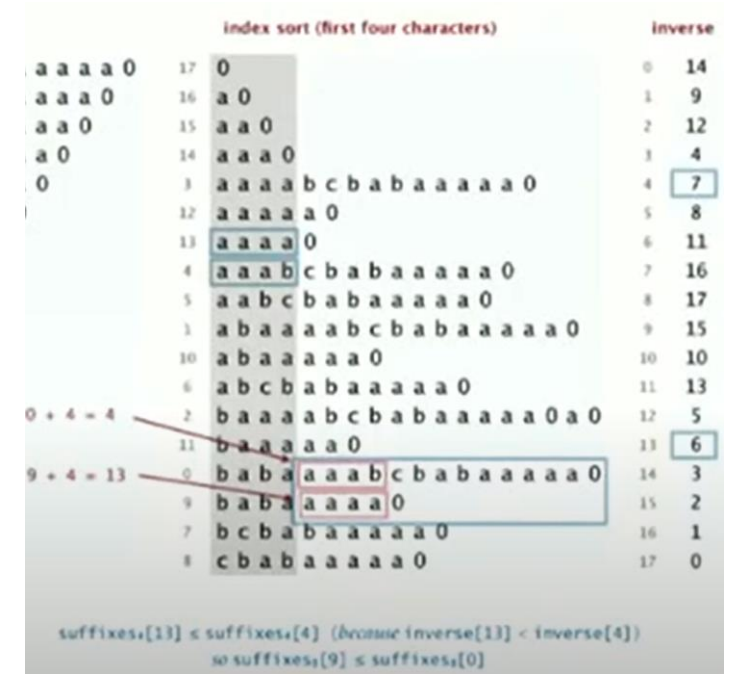
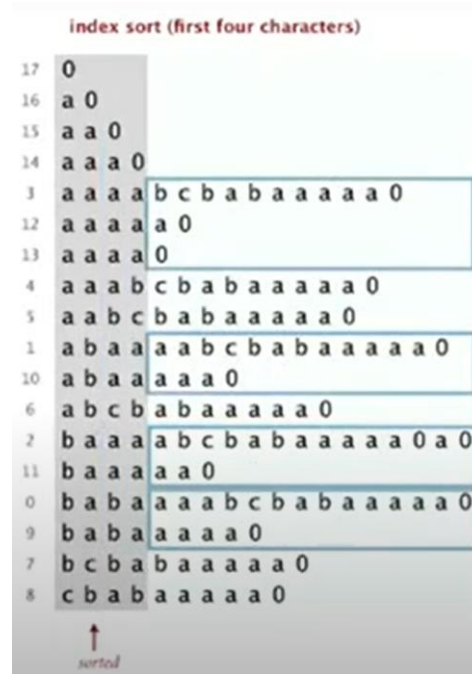
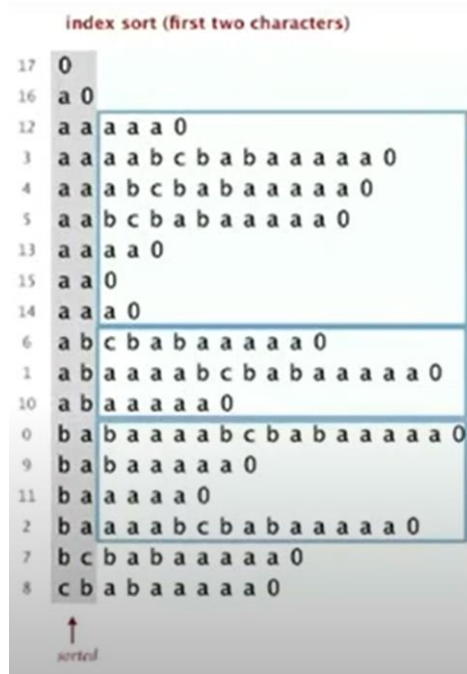
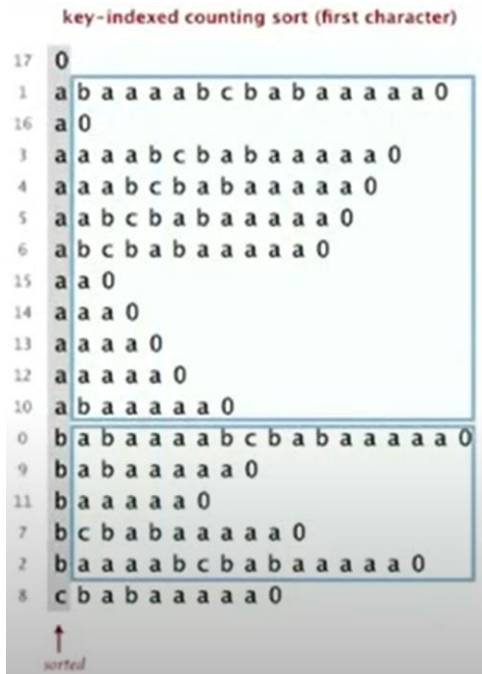


# Homework #1 Wrap-up

P<sub>10</sub> AAACCGTACACTGGGTTCAAGAGATTTCCC  
 P<sub>11</sub> AACCGTACACTGGGTTCAAGAGATTTCCC  
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 P<sub>31</sub> AGATTTCCC  
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 ⋮

## Variant algorithm(s)

- Sort phases take  $\log N$  because  $2^k$
- Each phase can be linear
- Limited benefit if not rate-limiting (i.e. cache misses)



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# Vectorized array operations in python with numpy

```
import numpy as np

# generate random data
x = np.random.randint(0, 10, size=(10, 3))
print(x)

[[9 1 8]
 [0 3 7]
 [4 7 4]
 [1 8 4]
 [2 7 2]
 [3 0 1]
 [7 5 7]
 [0 5 3]
 [0 8 1]
 [5 6 5]]

# sum down columns
np.sum(x, axis=0)

array([31, 50, 42])

# sum across rows
np.sum(x, axis=1)

array([18, 10, 15, 13, 11, 4, 19, 8, 9, 16])
```

```
# compute x^2
x ** 2

array([[81, 1, 64],
       [ 0, 9, 49],
       [16, 49, 16],
       [ 1, 64, 16],
       [ 4, 49, 4],
       [ 9, 0, 1],
       [49, 25, 49],
       [ 0, 25, 9],
       [ 0, 64, 1],
       [25, 36, 25]])
```

```
# make a binary mask for x == 7
(x == 7).astype(int)

array([[0, 0, 0],
       [0, 0, 1],
       [0, 1, 0],
       [0, 0, 0],
       [0, 1, 0],
       [0, 0, 0],
       [1, 0, 1],
       [0, 0, 0],
       [0, 0, 0],
       [0, 0, 0]])
```

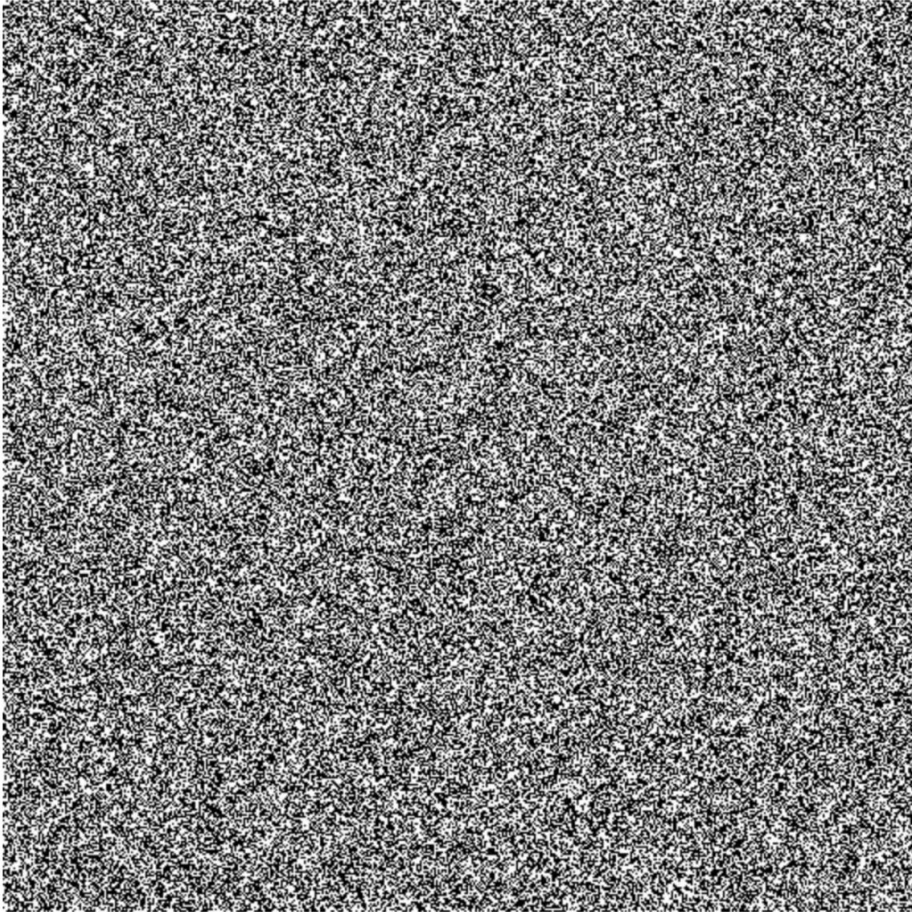
```
# transpose x
x.T

array([[9, 0, 4, 1, 2, 3, 7, 0, 0, 5],
       [1, 3, 7, 8, 7, 0, 5, 5, 8, 6],
       [8, 7, 4, 4, 2, 1, 7, 3, 1, 5]])
```

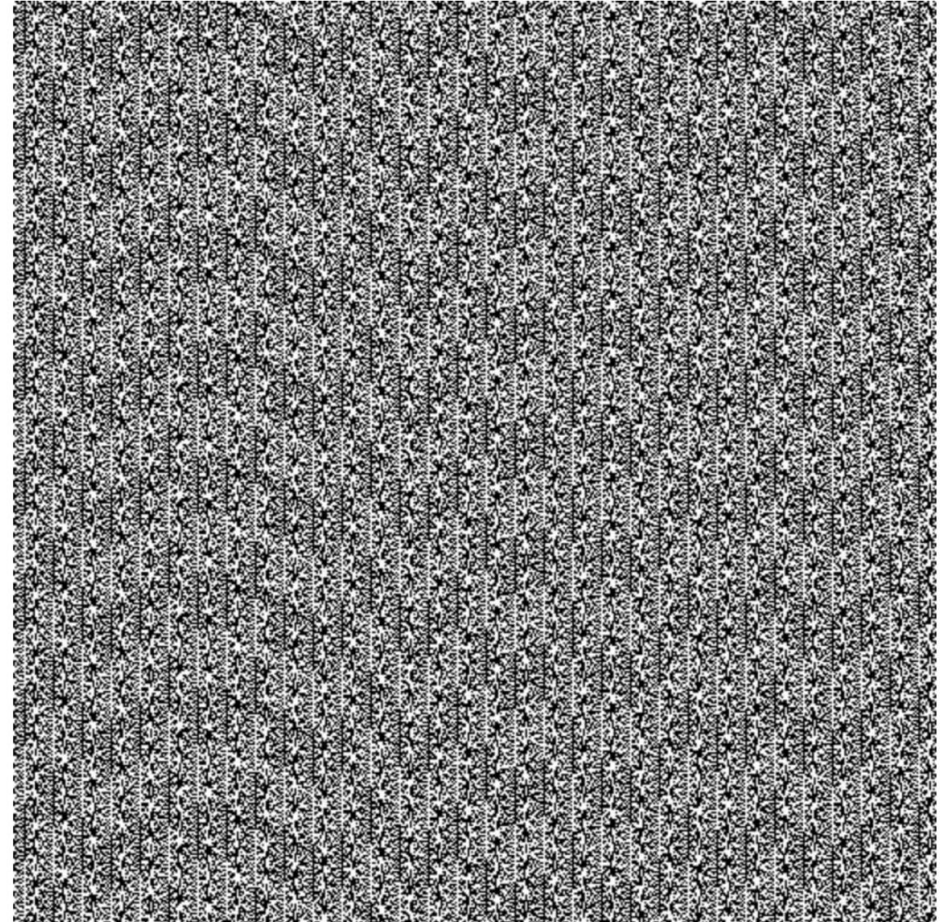


# Random number generation

C# System.Random



php rand()





# Vectorized array operations in python with numpy

```
import numpy as np

def vector_mult(x, c):
    '''Multiply all values in the input vector x by constant c.'''
    for i in range(len(x)):
        x[i] *= c
    return x

def vector_mult_np(x, c):
    '''Vectorized version of vector_mult using numpy.'''
    return x * c

# create a vector with some data
n = 10000 # length of vector to be multiplied
c = 25    # constant used in multiplication
x = list(range(n))
x_np = np.array(x)

# benchmark python list with iteration
%timeit vector_mult(list(x), 10)

# benchmark numpy vectorized calc
%timeit vector_mult_np(x_np.copy(), 10)

947 µs ± 272 ns per loop (mean ± std. dev. of 7 runs, 1000 loops each)
10.4 µs ± 175 ns per loop (mean ± std. dev. of 7 runs, 100000 loops each)
```

```
# example data
x = [0, 2, 4, 6, 8]
x_np = np.array([0, 2, 4, 6, 8])

vector_mult(list(x), 10)

[0, 20, 40, 60, 80]

vector_mult(list(x), 100)

[0, 200, 400, 600, 800]

vector_mult_np(x_np.copy(), 10)

array([ 0, 20, 40, 60, 80])

vector_mult_np(x_np.copy(), 100)

array([ 0, 200, 400, 600, 800])
```

# Random number generation in python

```
import random
```

```
x = [random.randint(0, 10) for i in range(5)]  
x
```

```
[2, 4, 4, 10, 4]
```

```
seq = [random.choice('ACGT') for i in range(10)]  
seq
```

```
['A', 'A', 'A', 'T', 'T', 'G', 'A', 'C', 'T', 'C']
```

```
import numpy as np
```

```
x = np.random.randint(0, 10, size=5)  
x
```

```
array([8, 8, 1, 0, 3])
```

```
seq = np.random.choice(['A', 'C', 'G', 'T'], size=10)  
seq
```

```
array(['T', 'G', 'G', 'A', 'T', 'A', 'A', 'C', 'A', 'G'], dtype='<U1')
```

```
seqs = np.random.choice(['A', 'C', 'G', 'T'], size=(5, 10))  
seqs
```

```
array([[ 'T', 'C', 'A', 'C', 'G', 'T', 'C', 'A', 'T', 'C'],  
      [ 'C', 'A', 'A', 'T', 'A', 'T', 'T', 'A', 'G', 'C'],  
      [ 'T', 'T', 'T', 'A', 'A', 'A', 'A', 'T', 'T', 'G'],  
      [ 'T', 'C', 'G', 'C', 'T', 'G', 'C', 'T', 'A', 'C'],  
      [ 'T', 'A', 'A', 'T', 'A', 'T', 'T', 'T', 'C', 'G']], dtype='<U1')
```

# Random number generation in C++

```
#include <iostream>
#include <cstdlib>
using namespace std;

int main()
{
    cout << "RAND_MAX:" << RAND_MAX << endl;
    for (int i = 0; i < 5; i++){
        cout << rand() << endl;
    }
}
```

rand() will return a random number between 0 and RAND\_MAX

```
RAND_MAX:2147483647
16807
282475249
1622650073
984943658
1144108930
RAND_MAX:2147483647
16807
282475249
1622650073
984943658
1144108930
RAND_MAX:2147483647
16807
282475249
1622650073
984943658
1144108930
```

Pseudo random number

# Random number generation in C++

```
#include <iostream>
#include <cstdlib>
#include <time.h>
using namespace std;

int main()
{
    cout << "RAND_MAX:" << RAND_MAX << endl;
    srand((unsigned)time(NULL));
    for (int i = 0; i < 5; i++){
        cout << rand() << endl;
    }
}
```

RAND\_MAX:2147483647

857283596

897610249

62834768

1649475099

861589770

RAND\_MAX:2147483647

857300403

1180085498

1685484841

486935110

2005698700

RAND\_MAX:2147483647

858241595

1966313913

212092108

1956688783

1651289370

```
cout << rand()/double(RAND_MAX) << endl; [0, 1]
```

```
cout << (rand() % (b-a+1)) + a << endl; [a, b]
```

# Random number generation in C++

## Using random() instead of rand()

```
#include <random>
#include <iostream>

int main()
{
    std::random_device rd;
    std::mt19937 mt(rd());
    std::uniform_real_distribution<double> dist(1.0, 10.0);

    for (int i=0; i<16; ++i)
        std::cout << dist(mt) << "\n";
}
```

rand() is typically a low quality pRNG. <random> provides a variety of engines with different characteristics suitable for many different use cases.

% generally biases the data and floating point division still produces non-uniform distributions. <random> distributions are higher quality as well as more readable.

srand() only permits a limited range of seeds. Engines in <random> can be initialized using seed sequences which permit the maximum possible seed data.



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# Homework 2 Overview

Part one: write a new program

- read in a file in FASTA format
- determine the frequencies of the nucleotides and dinucleotides (based on the forward strand) and the length of the sequence
- produce three simulated sequences based on the length and nucleotide or dinucleotide frequency of the original sequence
  - 'Equal frequency' model
  - Order 0 Markov model
  - Order 1 Markov model
- output three files in FASTA format containing the simulated sequence

# Homework 2 Overview

## order-0 Markov

### “Equal frequency” model

Nucleotide Frequencies:

A=0.2500

C=0.2500

G=0.2500

T=0.2500

```
Fasta 1: CP003913.fna
>gi|440453185|gb|CP003913.1|Mycoplasma pneumoniae M129-B7, complete genome
*=816373
A=249201
C=162924
G=163697
T=240551
N=0

Nucleotide Frequencies:
A=0.3053
C=0.1996
G=0.2005
T=0.2947
```

## order-1 Markov

	A	C	G	T
A	98512	50763	47914	52012
C	53047	36681	26746	46450
G	40870	37148	36764	48915
T	56772	38332	52273	93173

CGACTA

Dinucleotide Frequency Matrix:	
A=0.1207 0.0622 0.0587 0.0637	= 1
C=0.0650 0.0449 0.0328 0.0569	
G=0.0501 0.0455 0.0450 0.0599	
T=0.0695 0.0470 0.0640 0.1141	
Conditional Frequency Matrix:	
A=0.3953 0.2037 0.1923 0.2087	= 1
C=0.3256 0.2251 0.1642 0.2851	= 1
G=0.2497 0.2269 0.2246 0.2988	= 1
T=0.2360 0.1594 0.2173 0.3873	= 1

# Homework 2 Overview

Part two: run your HW1 program on three simulated genomes

- Run your HW1 program three times, using as input:
  - Human 10-Mb segment + simulated 'equal frequency' genome
  - Human 10-Mb segment + simulated Mouse Markov-0
  - Human 10-Mb segment + simulated Mouse Markov-1
- Given observed matches between the Human and simulated genomes, what can you conclude about the statistical significance of matches between the orthologous mouse and human regions in homework 1?

# Reminders

- Homework 2 due this Sunday Jan. 22, 11:59 pm
  - Single text file, compressed with `gzip`
  - name in the file: `camp_lisson_hw2.txt.gz`
- Homework 3 will be posted tomorrow (Jan. 18)



