

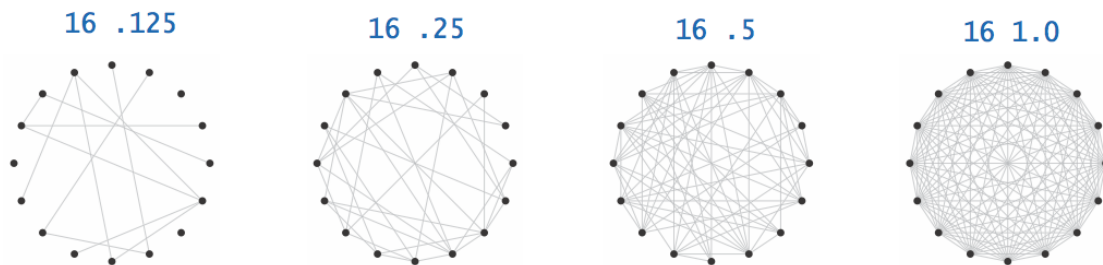
# Computer Science E214

## Tutorial 2

### 1 JBook 1.5.19

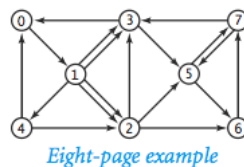
Write a program `randomgraph.py` that takes as command-line arguments an integer  $N$  and a floating-point value  $p \in [0,1]$ . The program should plot  $N$  equally spaced points on the circumference of a unit circle. Each pair of points should furthermore be connected by a line with probability  $p$ , i.e. the presence of each line is determined independently with probability  $p$ .

Example output for various parameters is given below.



### 2 Discerning Surfing (JBook 1.6.2, 1.6.11, and 1.6.12)

Encode the eight-page example shown below for use as input to `transition.py` in a file called `eightpage.txt`. Then implement and use `transition.py` to obtain a transition matrix for the example. Use this matrix to calculate the page ranks by implementing and using both `randomsurfer.py` and `markov.py`.



Now modify `transition.py` to ignore the effect of multiple links. That is, if there are multiple links from one page to another, count them as one link only. Call your modified program `singletransition.py`. Repeat the analysis above on the example, but using the modified transition matrix obtained with `singletransition.py`.

Repeat the analysis described above on the three-page example provided in `threepage.txt` in the tutorial resources.

Compare your results in each case: which do you think is a better approach to ranking web pages?

### 3 The binomial distribution (JBook 2.1.34)

Write a function

```
def binomial(n, k, p)
```

in a program `binomial.py` to compute the probability of obtaining exactly  $k$  heads in  $N$  flips of a biased coin (heads with probability  $p$ ) using the formula

$$f(n, k, p) = p^k (1 - p)^{n-k} N! / (k!(n - k)!)$$

*Hint:* To stave off overflow, compute  $x = \ln f(n, k, p)$  and then return  $e^x$ . You may want to use an additional function for this.

Write a `main` function for your program that obtains  $N$  and  $p$  from the command line and prints out

$$\sum_{k=0}^n f(n, k, p) ,$$

which should be equal to one (up to numerical accuracy). Set up your program to call the main function when it is executed, but not when it is imported: in the former case, the value of the special variable `__name__` is `'__main__'`; in the latter, it is the name of the module (i.e. `'binomial'` in this case).

Also use the class `binomialclient.py` provided in the tutorial resources to test your `binomial` function. Take a look at the contents of `binomialclient.py` and ensure you understand how it tests the code.

### 4 The Erlang loss formula

Consider a model of a telephone network switch where  $\lambda$  is the arrival rate of telephone calls (measured in calls per second) to a switch that can carry maximally  $N$  calls, and  $1/\mu$  is the average duration (measured in seconds) of a call. In telephony jargon,  $\lambda$  is called the *offered traffic* per unit time and  $\rho = \lambda/\mu$  the *offered load*. The load  $\rho$  is measured in units of “Erlangs”.

It is known that (under some reasonable assumptions we don’t go into here) the probability that an incoming call is lost because all  $N$  circuits are busy is a function of the load,  $\rho$ , and  $N$ . This *blocking probability* is given by the *Erlang loss formula* (also known as the *Erlang B-formula*):

$$B(N, \rho) = \frac{\rho^N / N!}{\sum_{n=0}^N \rho^n / n!} \quad (1)$$

Eqn. (1) looks easy to evaluate, but think of the situation where  $\rho = 100$  and  $N = 200$ . Then there is a term of the form  $(100)^{200}/200!$  that needs to be evaluated, and this is difficult to do exactly, even for a computer. To get around this we use a recursive formulation for  $B(N, \rho)$  which calculates  $B(N, \rho)$  in terms of  $B(N - 1, \rho)$

$$B(N, \rho) = \frac{\rho B(N - 1, \rho)}{N + \rho B(N - 1, \rho)} \quad (2)$$

where  $B(0, \rho) = 1$ .

Create a program `erlang.py`, and use it to perform the following:

1. Write a recursive function `recErlang` which takes two parameters  $n$  and  $\rho$  and computes and returns the blocking probability  $B(n, \rho)$  using Eqn. (2).

2. Determine the number of recursive calls that are used in total by this method to calculate the blocking probability — put your answer in the documentation for the method. (Note: If you are making more than one recursive call in your function, you are being inefficient, and you will run into trouble with large values of  $n$  — find a way to avoid it!)
3. Develop a non-recursive method `nonrecErlang` with the same parameters which also computes  $B(n, \rho)$ .
4. Implement a `main` function for testing that accepts  $n$  and  $\rho$  as command-line arguments, and then prints the results of the recursive and the non-recursive methods, as below:

```
$ python erlang.py 10 20
Recursive: B(10, 20.000000) = 0.537963
Non-recursive: B(10, 20.000000) = 0.537963
```

## 5 Recursive Squares (JBook 2.3.22)

Write a program `recursivesquares.py` to produce each of the following recursive patterns shown in Figures 1 - 4, using `stdraw` to draw the squares. The ratio of the sizes of the squares is 2.2 : 1. To draw a shaded square, draw a filled gray square, then an unfilled black square. Give the program two command-line parameters: the first should choose between the different recursive patterns: 1 for *Front*, 2 for *Order*, 3 for *Right*, or 4 for *Shaded*; the second value  $n$  should specify the depth of the recursion. Figures 1 - 4 show the results for recursion depths of *one* through *four* ((a) - (d)). An example for running the program:

```
$ python recursivesquares.py 1 4
```

should render Figure 1d.

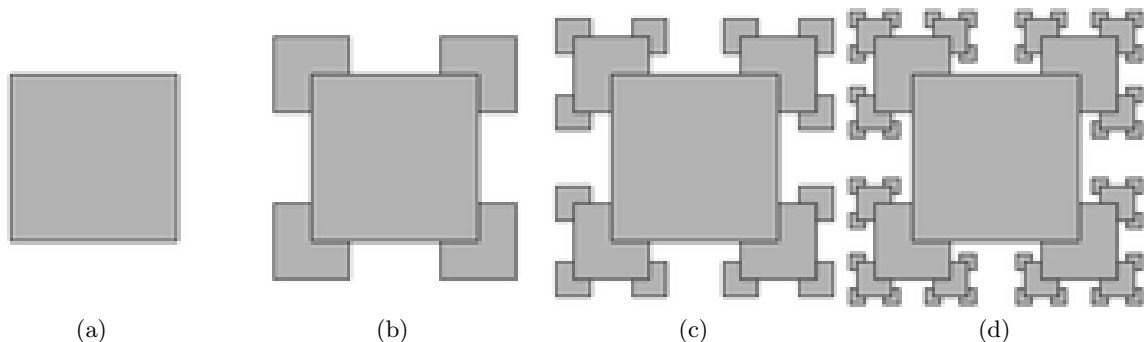


Figure 1: Front

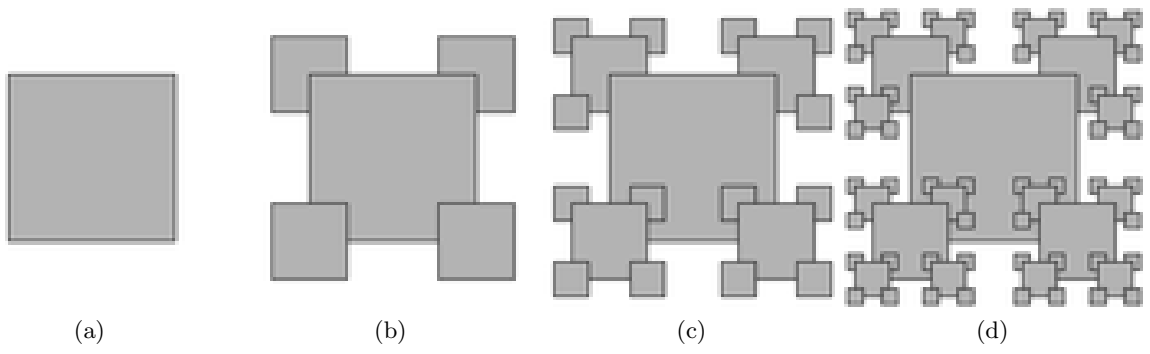


Figure 2: Order

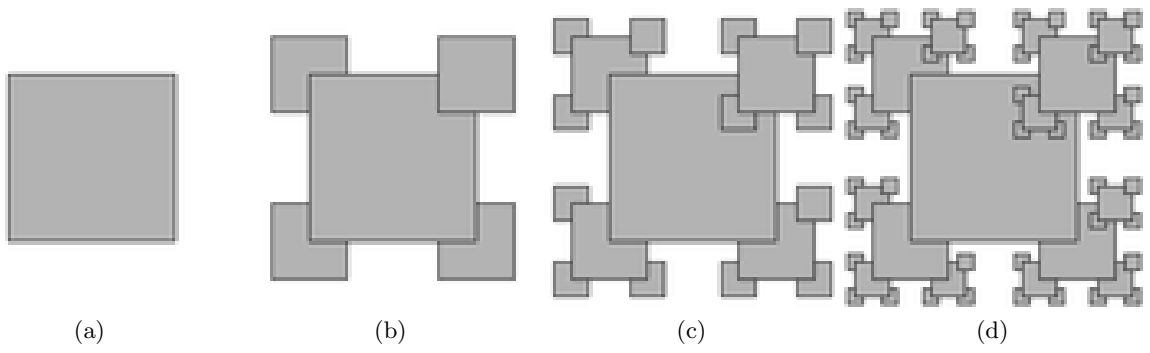


Figure 3: Right

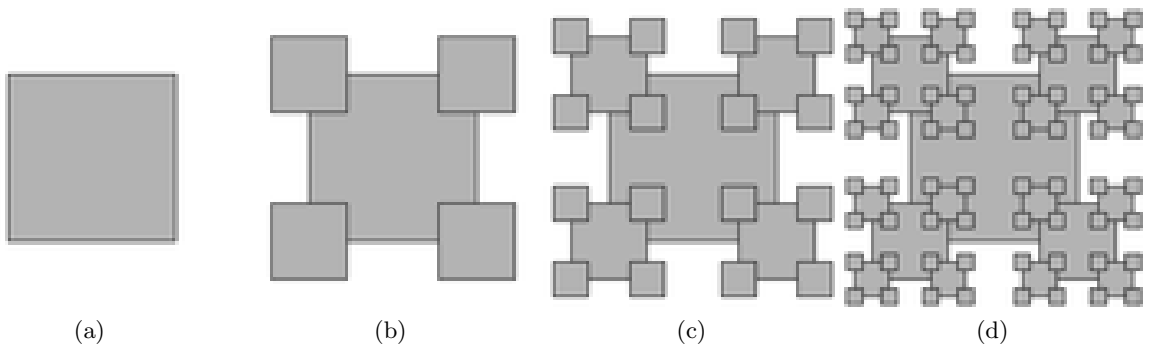


Figure 4: Shaded