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**Probability and statistics, 2021, Computer Science
Algorithmics,
Undergraduate Course, Part II, lecturer: Mirosław
Kutyłowski**

**I. Generating Random Numbers for
a given probability distribution**

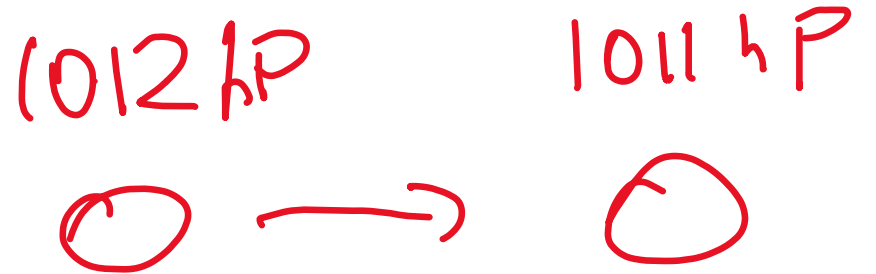
Chapter: 5.2 from Byron

goal:

**- simulations (e.g. pharma industry,
weather forecast, system testing ...)**

Weather simulations:

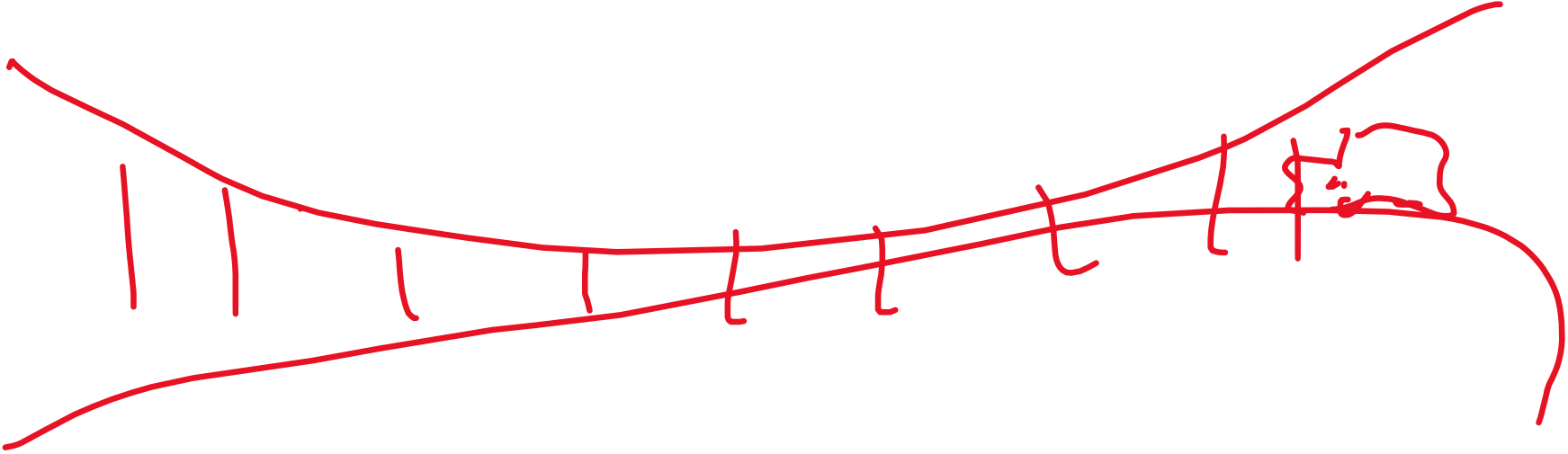
Stara metoda: modelovanie



Model: fluktuácie

Zloženy losový proces

Modelling behavior of a bridge:



The traffic on the bridge is a random proces: we need to generate these random events

Simulations for new chemical products, pharmaceuticals:

projekt. leków

Ziobonoski obliczeniowa

Physical sources: examples:

1) Electronics (bistable)



random()

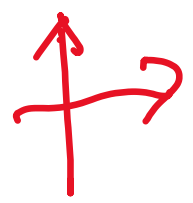
ring oscillator



3) quantum generators

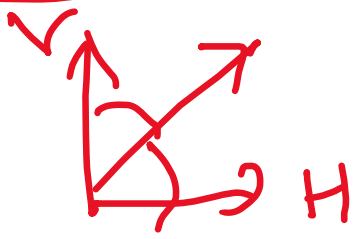
wysoka foton

read



3) noise

hp termiczny



Problems of physical sources:

1) bias

- obciążony
60% to 0, 40% bitów 1 | #0 ≠ #1

2) memory: dependence on history

ograniczenia na częstotliwości próbkowania

3) external influence

generowanie losowego hasła
zupat pseudydzalny

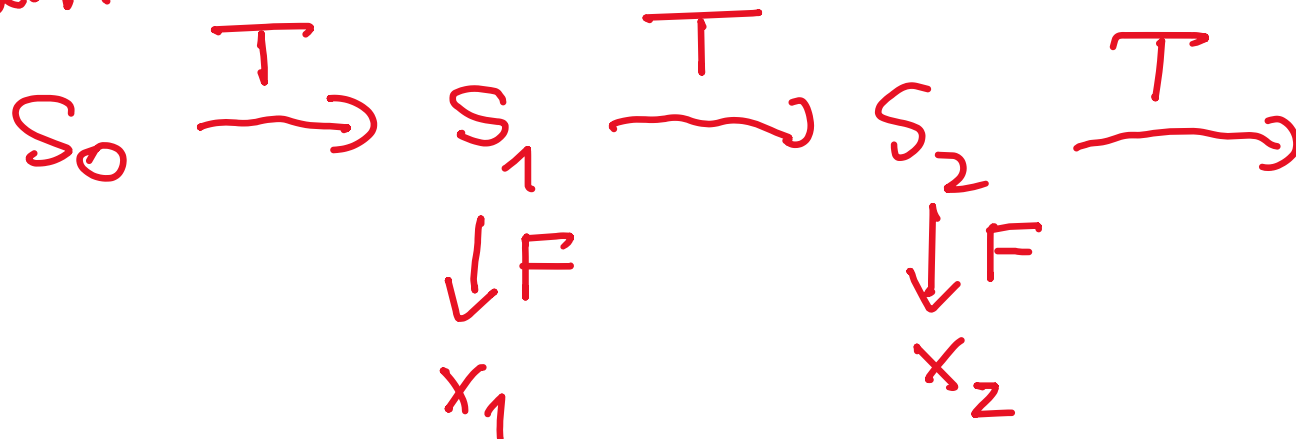
deterministic random number generators:

DRNG: NIST, recommendations,

rozkład jednostajny, ciąg bitów, „niezależny”

architecture: PRNG(seed) yields: bits

stan:



licznik:

licznik



DRNG:

basic property: not distinguishable from coin flipping

what does it mean?:

NIST tests

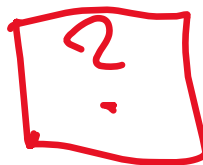
kompromis!

szybkie i w miarę skuteczne

left-or-right game



generator



A: DRNG



w środku dinozaur
bucza moneta;

secure PRNG: unpredictability

forwards:

Znając x_1, \dots, x_n nie mogą zgadnąć

x_{n+1}, x_{n+2}, \dots

backwards: szyfrowanie rozmów telefonicznych:

podstawiana rozmowa: $M \text{ XOR } S_1$

później: chwila milczenia: $0 \dots 0 \text{ XOR } S_2$

atakujący: zna S'

wyprowadza na podstawie S' ciąg S

odczytuje M

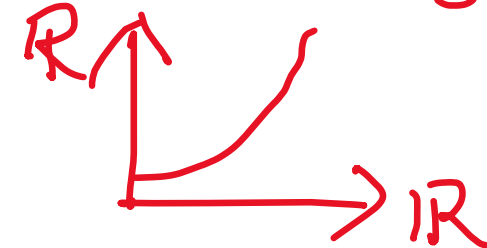
← tak nie powinno być!

realizations:

families of PRNG (based on residual arithmetic and algebraic expressions

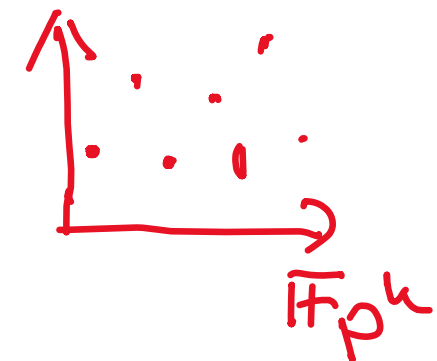
$$a \cdot x^2 + b \cdot x + c \pmod p$$

a, b - parametry



cryptographic generators: e.g. based on encryption

$$\text{trunc}(32, \text{Enc}_K(1)), \text{trunc}(32, \text{Enc}_K(2)), \dots$$

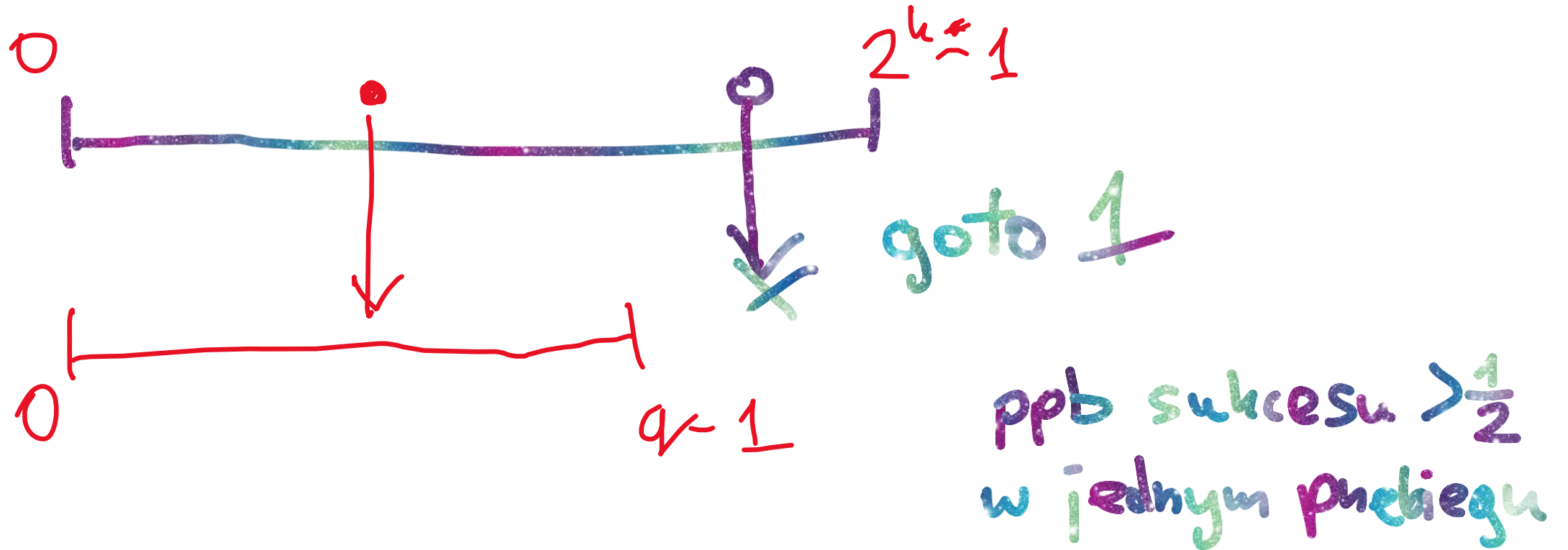


*bardzo efektywne gdy mamy np.
hardware'owe wsparcie dla szyfrowania
np.: AES*

Problem: domain

We have a good generator for uniform distribution over n bit numbers

How to get a uniform distribution over integers in the range $[0, q)$?



Problem: uniform versus non-uniform distribution

**all good PRNG resources deliver the output that is uniform over some interval
e.g.: 32 bit nonnegative integers**

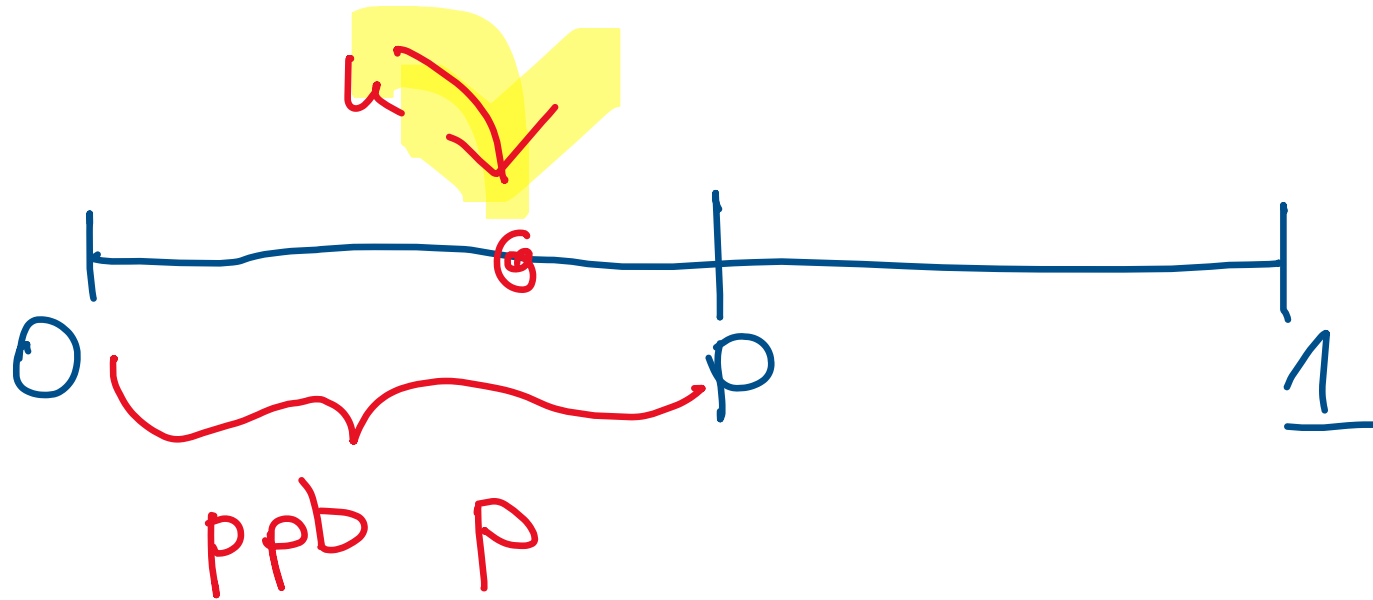
Needed:

e.g. geometric distribution, Poisson, ...

single Bernoulli trial:

procedure:

1. choose u uniformly at random in $[0,1]$
2. if $u < p$ then output 0 else output 1



1 bit:

$$0 \sim p \text{ pb } p^0$$

$$1 \sim p \text{ pb } 1-p$$

n Bernoulli trials, number of successes:

***n* times:**

0: with probability ***p***

1: with probability ***1-p***

count the number of successes

stupid solution:

compute pbb according to formulas

...

n Bernoulli trials, number of successes:

n times:

0: with probability p

1: with probability 1-p

count the number of successes

procedure (in MATLAB):

n=20; p=0.34;

U=rand(n,1); ←

X=sum(U<p)

↑
Summing over
the whole matrix

Matlab works on matrices. The function `rand(n,1)` generates a matrix $n \times 1$ with random elements from $[0,1)$

geometric distribution:

Bernoulli trials with pbb p of 0

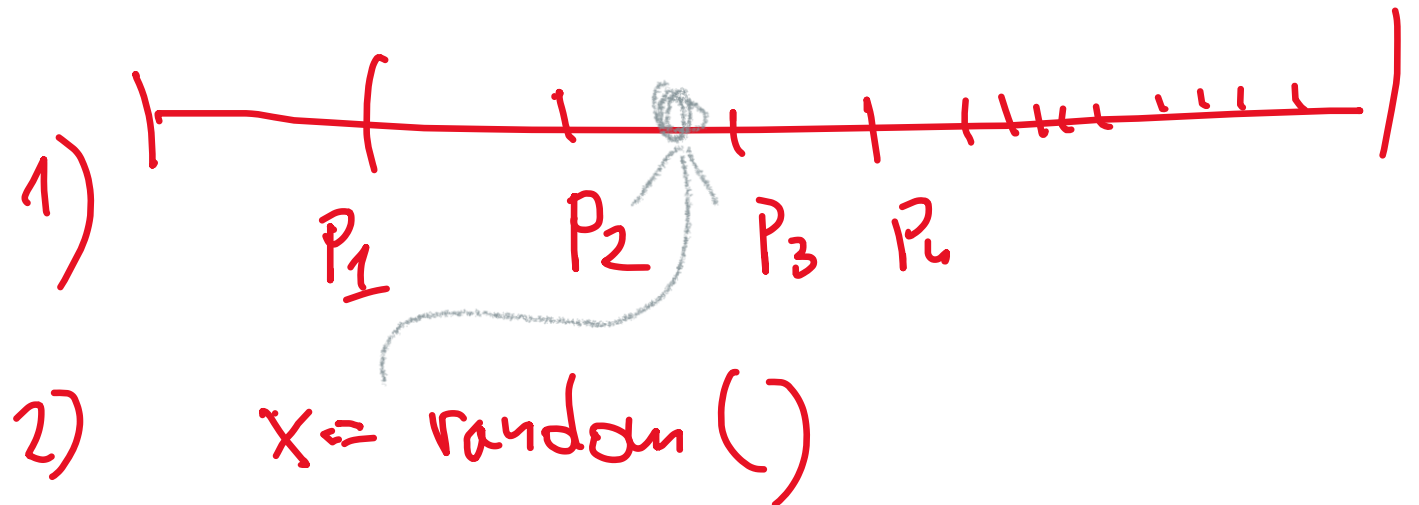
output: the number of trials until 1 chosen

naive way: take mathematical formulas and then choose according to the probabilities

procedure (in MATLAB):

```
X=1;  
while rand<p  
X=X+1;  
end;  
X
```

Naïve:



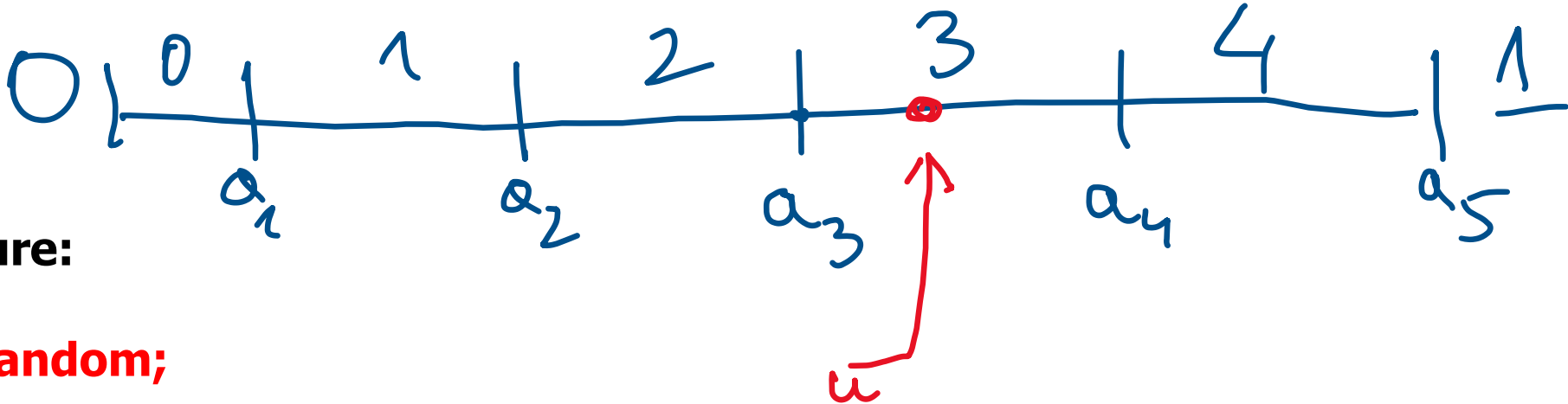
arbitrary discrete distribution:

assume: n possible values, $p(i)$ - probability of the i th value

approach: for each $i \leq n$ compute

$$a_i = \sum_{j < i} p_j$$

the results saved in a data structure D



procedure:

1. $u = \text{random};$
2. with D find i such that $a_i \leq u < a_{i+1}$

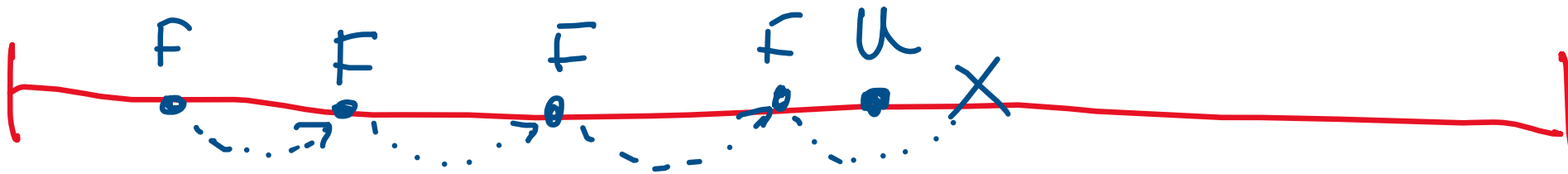
$$a_3 = p_{pb}(0) + p_{pb}(1) + p_{pb}(2)$$

Poisson distribution:

$$Pr(X=i) = e^{-\lambda} \cdot \frac{\lambda^i}{i!}$$

Matlab program from Byron:

```
lambda = 5;           % Parameter
U       = rand;       % Generated Uniform variable
i       = 0;         % Initial value
F       = exp(-lambda); % Initial value, F(0)
while (U >= F);      % The loop ends when U < F(i)
    F = F + exp(-lambda) * lambda^i / gamma(i+1);
    i = i + 1;
end;
X=i
```



tylko tyle iteracji ile trzeba

the case of invertible CDF

Theorem 2 Let X be a continuous random variable with cdf $F_X(x)$. Define a random variable $U = F_X(X)$. The distribution of U is $\text{Uniform}(0,1)$.

PROOF: First, we notice that $0 \leq F(x) \leq 1$ for all x , therefore, values of U lie in $[0, 1]$. Second, for any $u \in [0, 1]$, find the cdf of U ,

$$\begin{aligned} F_U(u) &= P\{U \leq u\} \\ &= P\{F_X(X) \leq u\} \\ &= P\{X \leq F_X^{-1}(u)\} && \text{(solve the inequality for } X\text{)} \\ &= F_X(F_X^{-1}(u)) && \text{(by definition of cdf)} \\ &= u && (F_X \text{ and } F_X^{-1} \text{ cancel)} \end{aligned}$$

