

# Computational Statistics Homework 1

Boubacar Sow

Exercise 1: Box-Muller and Marsaglia-Bray algorithm

let  $R$  a random variable  $\sim \text{Rayleigh}(1) \sim f_R(r) = r e^{(-\frac{r^2}{2})} \mathbb{1}_{R^+}(r)$ ,  $\forall r \in \mathbb{R}$

1. Let  $x$  and  $y$  such that:  $X = r \cos(\theta)$  and  $Y = r \sin(\theta)$

Prove that both  $X$  and  $Y \sim \mathcal{N}(0, 1)$  and are independant

We want to show that:  $E[h(X, Y)] = \int_{\mathbb{R}^2} h(x, y) f_X(x) f_Y(y) dx dy \quad \forall h \text{ continuous and bounded.}$

$$E[h(R \cos \theta, R \sin \theta)] = \int_{\mathbb{R}^2} h(r \cos \theta, r \sin \theta) f_{R, \Theta}(r, \theta) dr d\theta$$

Now we want to change variables from  $(r, \theta)$  to  $(x, y)$  using the transformation

$$\begin{aligned} \psi(r, \theta) &\mapsto (x, y) & \text{with } \psi \text{ a diffeomorphism: } \psi: \mathbb{R}^+ \times [0, 2\pi] \rightarrow \mathbb{R}^2 \\ \mathbb{R}^2 &\mapsto \mathbb{R}^2 & (r, \theta) \longrightarrow r \cos \theta, r \sin \theta \end{aligned}$$

$$\Rightarrow E[h(x, y)] = E[h(\psi(r, \theta))]$$

$$= \int_{\mathbb{R}^+} \int_0^{2\pi} h(\psi(r, \theta)) f_R(r) f_\Theta(\theta) dr d\theta$$

$$|\mathcal{J}_{\psi(r, \theta)}| = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r \cos^2 \theta + r \sin^2 \theta = r$$

$$\Rightarrow E[h(x, y)] = \iint_{\mathbb{R}^2} h(x, y) \frac{1}{2\pi} e^{-\frac{r^2}{2}} dx dy$$

$$\boxed{\begin{aligned} \text{We know that } x^2 + y^2 &= r^2 \\ &= \iint_{\mathbb{R}^2} h(x, y) e^{-\frac{(x^2+y^2)}{2}} \cdot \frac{1}{2\pi} dx dy = \iint_{\mathbb{R}^2} h(x, y) \underbrace{\frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}}_{f_X(x)} \underbrace{\frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}}_{f_Y(y)} dx dy \\ &= E[h(N_1, N_2)] \end{aligned}}$$

With  $N_1, N_2 \sim \mathcal{N}(0, 1)$  that are independant. We conclude that  $x, y \sim \mathcal{N}(0, 1)$ ,  $x \perp y$

2) Write an algorithm for sampling 2 independent gaussian distributions  $\mathcal{N}(0, 1)$

Cumulative distribution function (C.D.F):  $f_R(r) = \int_{-\infty}^r f_R(r) = P(R \leq r)$

let's  $U([0, 1])$ , then  $F_R^{-1}(u) \sim R$

Méthode de la transformée inverse

$F_x^{-1}(u)$ ,  $u \sim U([0, 1])$ , then  $F_x^{-1}(u) \sim X$

$$F_R(r) = -\int_0^r -r e^{-\frac{r^2}{2}} dr = [-e^{-\frac{r^2}{2}}] = 1 - e^{-\frac{r^2}{2}}$$

$$y = F_R(x) \Rightarrow y = 1 - e^{-\frac{x^2}{2}} \Rightarrow e^{-\frac{x^2}{2}} = 1 - y \text{ with } 0 < y < 1$$

$$\Rightarrow -\frac{x^2}{2} = \ln(1 - y) \quad y \sim U([0, 1])$$

$$\Rightarrow x = \sqrt{-2 \ln(1 - y)}$$

$$\Rightarrow F_R^{-1}(u) = \sqrt{-2 \ln(1 - u)}$$

Algorithm:

- Draw  $U_1 \sim U([0, 2\pi])$

$U_2 \sim U([0, 1])$

- Compute  $R = \sqrt{-2 \ln(1 - U_2)}$  where  $R = \sqrt{-2 \ln(U_2)}$  because  $1 - U_2 \sim U([0, 1])$

$$x = R \cos U_1 \sim \mathcal{N}(0, 1)$$

- Set  $\begin{cases} & , x, y \sim \mathcal{N}(0, 1) \text{ and } x \perp y \\ y = R \sin U_2 \sim \mathcal{N}(0, 1) & \end{cases}$

3. Consider the given Marsaglia-Bray algorithm

a- Distribution of  $(V_1, V_2)$  at the end of "while loop"

At the end of the while loop, we have  $V_1^2 + V_2^2 \leq 1$ . So  $V_1$  and  $V_2$  are on the disk with center being the origin and with radius = 1.

$V_1 = 2U_1 - 1$  and  $V_2 = 2U_2 - 1$ ;  $U_1$  and  $U_2 \sim U([0, 1])$  and  $U_1 \perp U_2 \Rightarrow V_1 \perp V_2$

Finally we conclude that  $(V_1, V_2) \sim U(D(0, 1))$

$$b. T_1 = \frac{V_1}{\sqrt{V_1^2 + V_2^2}} \quad T_2 = \frac{V_2}{\sqrt{V_1^2 + V_2^2}} \quad \text{and} \quad V = V_1^2 + V_2^2$$

Show that  $(T_1, T_2)$  and  $V$  are independant,  $V \sim U([0, 1])$  and  $(T_1, T_2)$  has the same distribution as  $(\cos(\theta), \sin(\theta))$  with  $\theta \sim U([0, 2\pi])$

$T_1 \perp\!\!\!\perp V$

let's  $h: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  a measurable, bounded function

$$\mathbb{E}[h(T_1, V)] = \mathbb{E}\left[h\left(\frac{V_1}{\sqrt{V_1^2 + V_2^2}}, (V_1^2 + V_2^2)\right)\right] = \iint h\left(\frac{V_1}{\sqrt{V_1^2 + V_2^2}}, (V_1^2 + V_2^2)\right) f_{V_1, V_2}(v_1, v_2) dv_1 dv_2$$

$$E = \{(v_1, v_2) \mid v_1^2 + v_2^2 \leq 1\}$$

$$f_{V_1, V_2}(v_1, v_2) = \frac{\mathbf{1}_{\{v_1^2 + v_2^2 \leq 1\}}}{\pi} \cdot \frac{1}{|E|} \quad |E| = \pi r^2 = \pi$$

Thus

$$\boxed{f_{V_1, V_2}(v_1, v_2) = \frac{\mathbf{1}_{\{v_1^2 + v_2^2 \leq 1\}}}{\pi} \cdot \frac{1}{\pi}}$$

$$\begin{aligned} \mathbb{E}[h(T_1, V)] &= \frac{1}{\pi} \iint_{V_1^2 + V_2^2 \leq 1} h\left(\frac{V_1}{\sqrt{V_1^2 + V_2^2}}, (V_1^2 + V_2^2)\right) dv_1 dv_2 \\ &= \frac{2}{\pi} \iint_{\{V_1^2 + V_2^2 \leq 1\} \cap \{V_2 > 0\}} h\left(\frac{V_1}{\sqrt{V_1^2 + V_2^2}}, (V_1^2 + V_2^2)\right) dv_1 dv_2 \end{aligned}$$

let's  $\phi: D(0, 1) \cap \{V_2 > 0\} \rightarrow [-1, 1] \times [0, 1]$

$$(V_1, V_2) \mapsto \left(\frac{V_1}{\sqrt{V_1^2 + V_2^2}}, V_1^2 + V_2^2\right)$$

$\phi$  is a diffeomorphism :  $\phi: D(0, 1) \cap \{V_2 > 0\} \rightarrow [-1, 1] \times [0, 1]$

$$|\mathcal{J}_\phi(v_1, v_2)| = \begin{vmatrix} \frac{v_2}{(v_1^2 + v_2^2)^{3/2}} & -\frac{v_1 v_2}{(v_1^2 + v_2^2)^{3/2}} \\ \frac{2v_1}{2v_1} & \frac{2v_2}{2v_2} \end{vmatrix}$$

$$\begin{aligned} &= \frac{2v_2^3 + 2v_1^2 v_2}{(v_1^2 + v_2^2)^{3/2}} = \frac{2v_2(v_1^2 + v_2^2)}{(v_1^2 + v_2^2)^{3/2}} = \frac{2v_2}{(v_1^2 + v_2^2)^{1/2}} \\ &= 2\sqrt{1 - \frac{v_1^2}{v_1^2 + v_2^2}} > 0 \end{aligned}$$

We have:

$$\mathbb{E}[h(T_1, V)] = \frac{1}{\pi} \int_{-1}^1 \int_0^1 h(t, v) \frac{1}{\sqrt{1-t^2}} dt dv$$

We can conclude that  $f_{T_1, V}(t, v) = \frac{1}{\pi\sqrt{1-t^2}} \mathbf{1}_{[-1, 1]}(t) \times \mathbf{1}_{[0, 1]}(v) \Rightarrow f_{V, T_1}(t, v) = f_{T_1}(t) \times f_V(v)$

$$\Rightarrow \boxed{T_1 \perp\!\!\!\perp V}$$

Independance of  $T_2$  and  $V$ :

$$\mathbb{E}[h(T_2, V)] = \frac{2}{\pi} \iint_{\{V_1^2 + V_2^2 \leq 1\} \cap \{V_1 > 0\}} h\left(\frac{V_2}{\sqrt{V_1^2 + V_2^2}}, (V_1^2 + V_2^2)\right) dv_1 dv_2$$

let's  $\phi: D(0, 1) \cap \{V_1 > 0\} \rightarrow [-1, 1] \times [0, 1]$  a diffeomorphism

$$(V_1, V_2) \mapsto \left(\frac{V_2}{\sqrt{V_1^2 + V_2^2}}, V_1^2 + V_2^2\right)$$

$$\text{The Jacobian is : } |\mathcal{J}_\phi(v_1, v_2)| = \begin{vmatrix} -\frac{v_1 v_2}{(v_1^2 + v_2^2)^{3/2}} & \frac{v_2^2}{(v_1^2 + v_2^2)^{3/2}} \\ \frac{2v_1}{2v_1} & \frac{2v_2}{2v_2} \end{vmatrix}$$

$$|\mathcal{J}_{\Phi(v_1, v_2)}| = 2 \sqrt{1 - \frac{v_1^2}{v_1^2 + v_2^2}} > 0$$

We have

$$\mathbb{E}[h(T_2, V)] = \frac{1}{\pi} \int_{-1}^1 \int_0^1 h(t, v) \frac{1}{\sqrt{1-t^2}} dt dv$$

Hence,  $T_2 \perp\!\!\!\perp V$  and  $\Rightarrow T_1, T_2 \perp\!\!\!\perp V$

let's show that  $(T_1, T_2)$  has the same distribution as:  $(\cos(\Theta), \sin(\Theta))$

let  $h$  a bounded measurable function, we have

$$\mathbb{E}[h(\cos(\Theta))] = \frac{1}{2\pi} \int_0^{2\pi} h(\cos(\theta)) d\theta = \frac{1}{2\pi} \int_0^\pi h(\cos(\theta)) d\theta + \int_\pi^{2\pi} h(\cos(\theta)) d\theta \quad (1)$$

$$(2)$$

$$(1): \int_0^\pi h(\cos(\theta)) d\theta = - \int_{-1}^1 h(t) \sin(\theta) dt = \int_{-1}^1 h(t) \frac{1}{\sqrt{1-t^2}} dt$$

With a change of variable  $t = \cos(\theta)$  and  $dt = -\sin(\theta) d\theta$

$$t^2 = \cos^2(\theta) \Rightarrow t^2 = 1 - \sin^2(\theta) \Rightarrow \sin(\theta) = \sqrt{1-t^2} \quad w = (1, 2)$$

$$x-i = (1, 2)$$

$$(2): \int_\pi^{2\pi} h(\cos(\theta)) d\theta = \int_0^\pi h(\cos(u+\pi)) du = \int_0^\pi h(-\cos(u)) du \quad (1, 2) \times (1, 1) - (1, 2)(2, 2)$$

with a change of variable  $u = \theta - \pi$ ,  $du = d\theta$

let's  $t = -\cos(u)$  and  $dt = \sin(u) du = \sqrt{1-t^2} du$

$$\Rightarrow \int_{-1}^1 h(t) \frac{1}{\sqrt{1-t^2}} dt$$

$$\text{Hence: } \mathbb{E}[h(\cos(\Theta))] = \frac{1}{2\pi} \left[ 2 \int_{-1}^1 h(t) \frac{1}{\sqrt{1-t^2}} dt \right]$$

$$= \int_{-1}^1 h(t) \frac{1}{\pi \sqrt{1-t^2}} dt$$

We have here  $f_{T_1}(t) = \frac{1}{\pi \sqrt{1-t^2}} \mathbf{1}_{[-1,1]}$ , thus we conclude that  $T_1 \sim \cos(\Theta)$  and  $V \sim \mathcal{U}(0,1)$

By the same procedure we can get  $T_2 \sim \sin(\Theta)$ .

Finally  $(T_1, T_2) \sim (\cos(\Theta), \sin(\Theta))$

c) Distribution of the output  $(X, Y)$

Dans l'algorithme :

$$S = \sqrt{-2 \log r} = F_R^{-1}(1-v) \quad \text{with } V \sim \mathcal{U}(0,1)$$

Thus  $1-v \sim \mathcal{U}(0,1) \Rightarrow S \sim \text{Rayleigh}(1)$

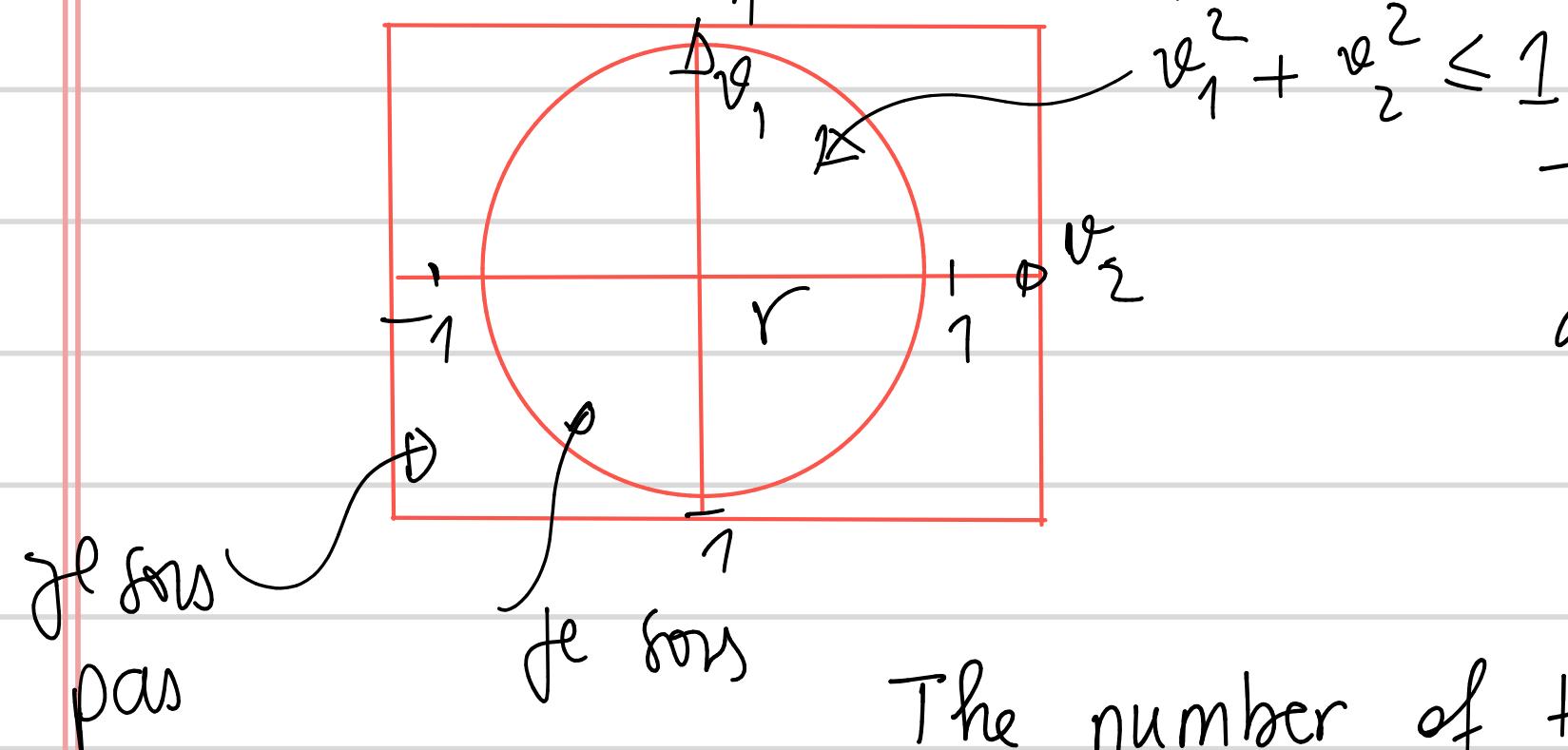
We can also remark that:

$$X = ST_1 \text{ and } Y = ST_2 \quad \text{with } (T_1, T_2) \sim (\cos(\Theta), \sin(\Theta)), \Theta \sim \mathcal{U}(0,1)$$

$V \perp\!\!\!\perp (T_1, T_2)$  thus  $S \perp\!\!\!\perp \Theta$

From question (1), we know that  $X \perp\!\!\!\perp Y$ , we conclude that  $X, Y \stackrel{i.i.d.}{\sim} \mathcal{N}(0,1)$

d) Expected number of steps in the while loop



The probability that we are in the circle given that we are in  $[-1, 1]^2$  is

$$P = \frac{|E|}{|C|} = \frac{\pi r^2}{4r^2} = \frac{\pi}{4}$$

The number of times we will go through the loop follows a geometric law

$$P(s = n) = \left(1 - \frac{\pi}{4}\right)^{n-1} \frac{\pi}{4}$$

stay in the loop      get out the while loop

It is thus the expected value of a geometric law with parameter  $p = \frac{\pi}{4}$   
 $\Rightarrow N \sim \text{Geometric}\left(\frac{\pi}{4}\right)$

$$\text{Number} = E[N] = \frac{1}{p} = \frac{4}{\pi} \approx 2$$

The number of times we will go through the while loop is : 2

### Exercise 2: Invariant distribution

We define a Markov chain  $(X_n)_{n \geq 0}$  in  $[0, 1]$ , ---

If  $X_n = \frac{1}{m}$  ( $\forall m > 0$ ), we let:

$$\begin{cases} X_{n+1} = \frac{1}{m+1} \text{ with probability } 1 - x_n^2 \\ X_{n+1} \sim U([0, 1]) \text{ with probability } x_n^2 \end{cases}$$

If not,  $X_{n+1} \sim U([0, 1])$

1. Prove that the transition kernel of the chain  $(X_n)_{n \geq 0}$  is given by:

$$P(x, A) = \begin{cases} x^2 \int_{A \cap [0, 1]} dt + (1-x^2) \delta_{\left\{\frac{1}{m+1}\right\}}(A) & \text{if } x = \frac{1}{m} \\ \int_{A \cap [0, 1]} dt & \text{otherwise} \end{cases}$$

where  $\delta_d$  is the Dirac measure at  $d$

let's  $Q = \left\{ \frac{1}{m}, m \in \mathbb{N}^* \right\}$ , transition kernel = distribution  $(X_{n+1} | X_n)$

$$\mathbb{E}[h(X_{n+1} | X_n)] = \int h(y) \underbrace{P(x, dy)}_{\substack{\text{"density of } X_{n+1} | X_n}} \quad \forall h \text{ bounded and continuous}$$

let's  $h$  a continuous and bounded function, we have:  $\mathbb{E}[h(X_{n+1}) | X_n = x_n] = \int h(y) P(x, dy)$

If  $x \in Q$

$$\mathbb{E}[h(X_{n+1} | X_n = x)] = \int_{\mathbb{R}} h(y) \underbrace{P(x, dy)}_{f_{X_{n+1} | X_n}(y) dy} = \int_{\mathbb{R}} h(y) \underbrace{\mathbb{1}_{[0, 1]}(y) dy}_{P(x, dy) \text{ does not depend on } x} \quad \checkmark x \notin Q$$

$$\implies P(x, A) = \int_A P(x, dy) = \int_{A \cap [0, 1]} dy$$

If  $x \in Q$ ,  $B_m$  is a random Bernoulli variable, with parameter  $\frac{1}{m^n} \left[ \equiv B_m \sim \mathcal{B}\left(\frac{1}{m^n}\right) \right]$

$$\mathbb{E}[h(X_{n+1} | X_n = x)] = \mathbb{E}[h(X_{n+1}) | X_n = x, B_m = 0] P(B_m = 0) + \mathbb{E}[h(X_{n+1}) | X_n = x, B_m = 1] P(B_m = 1)$$

$$= h\left(\frac{1}{m+1}\right) \left(1 - \frac{1}{m^2}\right) + \frac{1}{m} \int_{[0, 1]} h(y) dy$$

$$\implies P(x, A) = (1-x^2) \delta_{\frac{1}{m+1}}(A) + x^2 \int_{[0, 1]} h(y) dy \quad \text{with } x = \frac{1}{m}$$

$$\text{We have: } P(x, A) = \begin{cases} x^2 \int_{A \cap [0, 1]} dt + (1-x^2) \delta_{\left\{\frac{1}{m+1}\right\}}(A) & \text{if } x = \frac{1}{m} \\ \int_{A \cap [0, 1]} dt & \text{otherwise} \end{cases}$$

2. Prove that uniform distribution on  $[0, 1]$  is invariant for  $P$ . In the following, this invariant distribution will be denoted by  $\pi$

A distribution  $\pi$  is invariant for a Markov chain with transition kernel  $P$  if for every measurable set  $A$ , we have:  $\pi(A) = \int_{[0, 1]} P(x, A) \pi(dx) \quad [\pi(A) = \pi P(A)]$

We want to show that the uniform distribution on  $[0, 1]$  is invariant for  $P$ .

$$\pi P(A) = \int_{\mathbb{R}} P(x, A) \pi(dx) = \int_Q P(x, A) \pi(dx) + \int_{\bar{Q}} P(x, A) \pi(dx)$$

(1) = 0, because it's over a countable set of points [the points  $x = \frac{1}{m}$ , for  $m = 1, 2, \dots$ ]

$$(2) = \int_{A \cap [0, 1]} dt \pi(dx)$$

$$\implies \pi P(A) = \int_{\bar{Q}} \int_{A \cap [0, 1]} dt \pi(dx) = \int_{A \cap [0, 1]} \int_{\bar{Q}} \pi(dx)$$

$$\Rightarrow \Pi P(A) = \int_{\bar{Q}} \int_{A \cap [0,1]} dt \Pi(dx) = \underbrace{\int_{A \cap [0,1]} \int_{\bar{Q}} \Pi(dx)}_{\Pi(A)} \underbrace{1}_{1}$$

$$= \Pi(A)$$

3. Let  $x \notin \{\frac{1}{m}, m \in \mathbb{N}^*\}$ . Compute the value  $P_f(x) = \mathbb{E}[f(X_1) | X_0 = x]$  for a bounded measurable function  $f$ .

$\downarrow X_1$  is uniformly distributed in  $[0,1]$

$$P_f(x) = \mathbb{E}[f(x_1) | X_0 = x] = \int_0^1 f(u) P(x, du) = \int_0^1 f(u) du$$

Deduce  $P^n f(x) \quad \forall n \geq 1$ :

$$P^1 f(x) = P_f(x) = \Pi(f)$$

$$P^2 f(x) = \int_0^1 \int_0^1 f(u) du dv = \int_0^1 f(u) du$$

$$\text{let's assume } P^n f(x) = \int_0^1 f(u) du \quad \forall n \geq 1$$

We need to show that the statement is true for  $n+1$ , i.e

$$P^{n+1} f(x) = \int_0^1 f(u) du$$

$$\begin{aligned} P^{n+1} f(x) &= P(P^n(f(x))) = \int_{\bar{A}} P^n f(v) P(x, dv) = \int_0^1 P^n f(v) dv \\ &= \int_0^1 \int_0^1 f(u) du dv = \int_0^1 f(u) du \end{aligned}$$

$$\text{Thus we conclude that } P^n f(x) = \int_0^1 f(u) du \quad \forall n \geq 1$$

4. let  $x = \frac{1}{m}$  with  $m \geq 2$

a- Let  $n \in \mathbb{N}$ . Compute  $P^n(x, \frac{1}{n+m})$  in terms of  $m$  and  $n$

If  $n=1$ :

$$\begin{aligned} P(x, \frac{1}{1+m}) &= P\left(\frac{1}{m}, \frac{1}{m+1}\right) \\ &= P(X_{n+1} = \frac{1}{m+1} | X_n = \frac{1}{m}) \end{aligned}$$

$$P(x, \frac{1}{m+1}) = 1 - x^2 = 1 - \left(\frac{1}{m}\right)^2$$

For  $n=2$ :

$$\begin{aligned} P(x, \frac{1}{m+2}) &= P\left(P\left(\frac{1}{m}, \frac{1}{m+1}\right)\right) = \int P(y, \frac{1}{m+2}) P(x, dy) \\ &= \int P(y, \frac{1}{m+2}) \left[ x^2 \int_{dy \cap [0,1]} du + (1-x^2) \delta_{\frac{1}{m+1}} dy \right] \\ &= \int P(y, \frac{1}{m+2}) (1-x^2) \delta_{\frac{1}{m+1}} dy = (1-x^2) \int P(y, \frac{1}{m+2}) \delta_{\frac{1}{m+1}} dy \\ &= \left(1 - \left(\frac{1}{m}\right)^2\right) \int P(y, \frac{1}{m+2}) \delta_{\frac{1}{m+1}} dy = \left(1 - \frac{1}{m^2}\right) P\left(\frac{1}{m+1}, \frac{1}{m+2}\right) \\ &= \left(1 - \left(\frac{1}{m}\right)^2\right) \left(1 - \left(\frac{1}{m+1}\right)^2\right) = \prod_{i=0}^1 \left(1 - \left(\frac{1}{m+i}\right)^2\right) \end{aligned}$$

$$P^2(x, \frac{1}{m+2}) = \prod_{i=0}^2 (1 - (\frac{1}{m+i})^2)$$

Heredite' :

$$\text{Let's assume for any } k \text{ we have : } P^k(x, \frac{1}{m+k}) = \prod_{i=0}^{k-1} (1 - (\frac{1}{m+i})^2)$$

$$\text{let's try to show that we have for } k+1 : P^{k+1}(x, \frac{1}{m+k+1}) = \prod_{i=0}^k (1 - (\frac{1}{m+i})^2)$$

$$P^{k+1}(x, \frac{1}{m+k+1}) = P(P^k(x, \frac{1}{m+k+1}))$$

$$= \int P^k(y, \frac{1}{m+k+1}) P(x, dy) = \int P^k(y, \frac{1}{m+k+1}) \left[ x^2 \int_{dy \in [0, 1]} du + (1-x^2) \delta_{\frac{1}{m+1}} dy \right]$$

$$= (1 - (\frac{1}{m})^2) P^k(\frac{1}{m+1}, \frac{1}{m+k+1})$$

Taking  $m' = m+1$ , we have

$$P^{k+1}(x, \frac{1}{m+k+1}) = (1 - (\frac{1}{m})^2) P^k(\frac{1}{m}, \frac{1}{m'+k})$$

Using our hypothesis on  $P^k(x, \frac{1}{m+k})$ , we have

$$P^{k+1}(x, \frac{1}{m+k+1}) = (1 - (\frac{1}{m})^2) \prod_{i=0}^{k-1} (1 - (\frac{1}{m'+i})^2)$$

$$= (1 - (\frac{1}{m})^2) \prod_{i=0}^{k-1} (1 - (\frac{1}{m+i+1})^2)$$

$$= (1 - (\frac{1}{m})^2) \prod_{i=0}^k (1 - (\frac{1}{m+i})^2)$$

$$P^{k+1}(x, \frac{1}{m+k+1}) = \prod_{i=0}^k (1 - (\frac{1}{m+i})^2)$$

We have shown that  $P^n(x, \frac{1}{m+n}) = \prod_{i=0}^{n-1} (1 - (\frac{1}{m+i})^2)$

b) Do we have  $\lim_{n \rightarrow +\infty} P^n(x, A) = \pi(A)$  when  $A = \bigcup_{q \in \mathbb{N}} \{\frac{1}{m+1+q}\}$

$$\pi(A) = \sum_{q \in \mathbb{N}} \pi\left(\left\{\frac{1}{m+1+q}\right\}\right) = 0$$

$\forall n \in \mathbb{N}_+^*$ , we have:

$$P^n(x, A) = \sum_{q \in \mathbb{N}} P^n\left(\frac{1}{m}, \frac{1}{m+q+1}\right) = P^n\left(\frac{1}{m}, \frac{1}{m+n}\right)$$

$$= \prod_{i=0}^{n-1} \left(1 - \left(\frac{1}{m+i}\right)^2\right) = \prod_{i=0}^{n-1} \left(\frac{(m+i)^2 - 1}{(m+i)^2}\right)$$

$$= \prod_{i=0}^{n-1} \left(\frac{(m+i+1)(m+i-1)}{(m+i)^2}\right) = \frac{\prod_{i=0}^{n-1} (m+i-1) \cdot \prod_{i=0}^{n-1} (m+i+1)}{\prod_{i=0}^{n-1} (m+i)^2}$$

$$= \frac{\prod_{i=1}^{n-2} (m+i) \times \prod_{i=1}^n (m+i)}{\prod_{i=0}^{n-1} (m+i)^2} = \frac{(m-1)(m+n)}{m(m+n-1)} = \frac{m-1}{m} \left(1 + \frac{1}{m+n-1}\right)$$

$$\geq \frac{1}{2} > \pi(A) = 0$$

We conclude that  $\lim_{n \rightarrow +\infty} P^n(x, A) \neq \pi(A)$