

Vitesse de Convergence d'une chaîne de Markov ergodique vers sa mesure stationnaire

```
> restart():
with(linalg):
with(stats):
with(LinearAlgebra):
Randomize():
with(RandomTools):
with(plots):
Warning, the previous binding of the name GramSchmidt has been removed and it
now has an assigned value

Warning, these names have been redefined: anova, describe, fit, importdata,
random, statevalf, statplots, transform

Warning, the name changecoords has been redefined
```

– génération aléatoire d'une matrice stochastique (matrice de transition de la chaîne)

– la petite fonction...

```
> rand_stoch := proc (n)
local v,i,j,s;
v := Matrix(n,n);
for j from 1 to n do
s := 0;
for i from 1 to n do
v[j,i] := rand();
s := s + v[j,i];
end do;
for i from 1 to n do
v[j,i] := v[j,i]/s;
end do;
end do;
v;
end proc;

rand_stoch := proc(n)
local v, i, j, s;
v := Matrix(n, n);
for j to n do
s := 0;
for i to n do v[j, i] := rand( ); s := s + v[j, i] end do;
for i to n do v[j, i] := v[j, i] / s end do
end do;
v
```

```
end proc
```

... et son application

```
> Q := rand_stoch(4):  
evalf(Q);  
Id:=Matrix(4,4,shape=identity):  

$$\begin{bmatrix} 0.5206840165 & 0.1642503752 & 0.3022593487 & 0.01280625953 \\ 0.1744774040 & 0.07826179288 & 0.6826583913 & 0.06460241180 \\ 0.1881386595 & 0.2799814644 & 0.3290417997 & 0.2028380765 \\ 0.4572083766 & 0.08482520852 & 0.1054323232 & 0.3525340917 \end{bmatrix}$$

```

calcul de sa loi invariante

résolution du système

```
> M := Transpose(Id-Q):  
pi_ := LinearSolve(M, Vector(4,0), free='s');  

$$pi_ := \begin{bmatrix} \frac{1061428513435098349023713888023092680658852058727}{431045428304731078196580268678806113629775671320} s_4 \\ \frac{71392423954001540454837701810250136593193740793}{53880678538091384774572533584850764203721958915} s_4 \\ \frac{93916172177331839759517244383417767757008961567}{35920452358727589849715022389900509469147972610} s_4 \\ s_4 \end{bmatrix}$$

```

normalisation de la loi invariante

```
> i := 'i':  
x := solve(sum('pi_[i]', i=1..4)=1, {'s[1]', 's[2]',  
's[3]', 's[4]'}):  
pi := Transpose(subs(x, pi_)):  
evalf(pi);  

$$[0.3326728677, 0.1790064775, 0.3532224198, 0.1350982350]$$

```

Tirage d'une loi sur un ensemble à quatre éléments (pour initialiser la chaîne)

```
> a := rand():  
b := rand():  
c := rand():  
d := rand():
```

```

S := a+b+c+d:
t_max := 15;
mu := Matrix(4, t_max+1, 0):
mu[1,1] := a/S:
mu[2,1] := b/S:
mu[3,1] := c/S:
mu[4,1] := d/S:
evalf(Column(mu,1));

```

```

t_max := 15
[ 0.4151179114
  0.07188356699
  0.3938522326
  0.1191462890 ]

```

- calcul de la loi itérée k fois

```

> for t from 1 to t_max do
  eta[t] :=
  VectorMatrixMultiply(Transpose(Column(mu,1)), MatrixPower(Q,t)
  ):
end do:
evalf(eta[t_max]);

```

[0.1986515648, 0.2080541973, 0.3508591358, 0.2424351021]

- Calcul de N

```

> d :=2:
f:= (i,j) -> 4*(eta[j][i] - eta[j][i]^2)*10^(2*d):
N_mat := evalf(Matrix(4,t_max, f)):
N := floor(max(seq(seq(N_mat[i,j], j = 1 .. t_max ), i = 1
.. 4)));

```

N := 9379

- simulation de la loi invariante - mesures empiriques

```

> min(seq(seq(Q[i,j], j = 1 .. 4 ), i = 1 .. 4)):N_inv := 1/N:
for t from 2 to t_max+1 do
  M := MatrixPower(Q,t-1):
  p := VectorMatrixMultiply(Transpose(Column(mu,1)), M);
  p2 := p[1]+p[2]:
  p3 := p2+p[3]:
  for i from 1 to N do
    g := Generate(float(method=uniform)):
    if g < p[1] then
      mu[1,t] := mu[1,t]+N_inv;
    elif g < p2 then
      mu[2,t] := mu[2,t]+N_inv;
    elif g < p3 then

```

```

        mu[3,t] := mu[3,t]+N_inv;
    else
        mu[4,t] := mu[4,t]+N_inv;

    end if;
end do:
end do:
evalf(Column(mu,t_max+1));

```

```

[0.1974624160]
[0.2067384583]
[0.3543021644]
[0.2414969613]

```

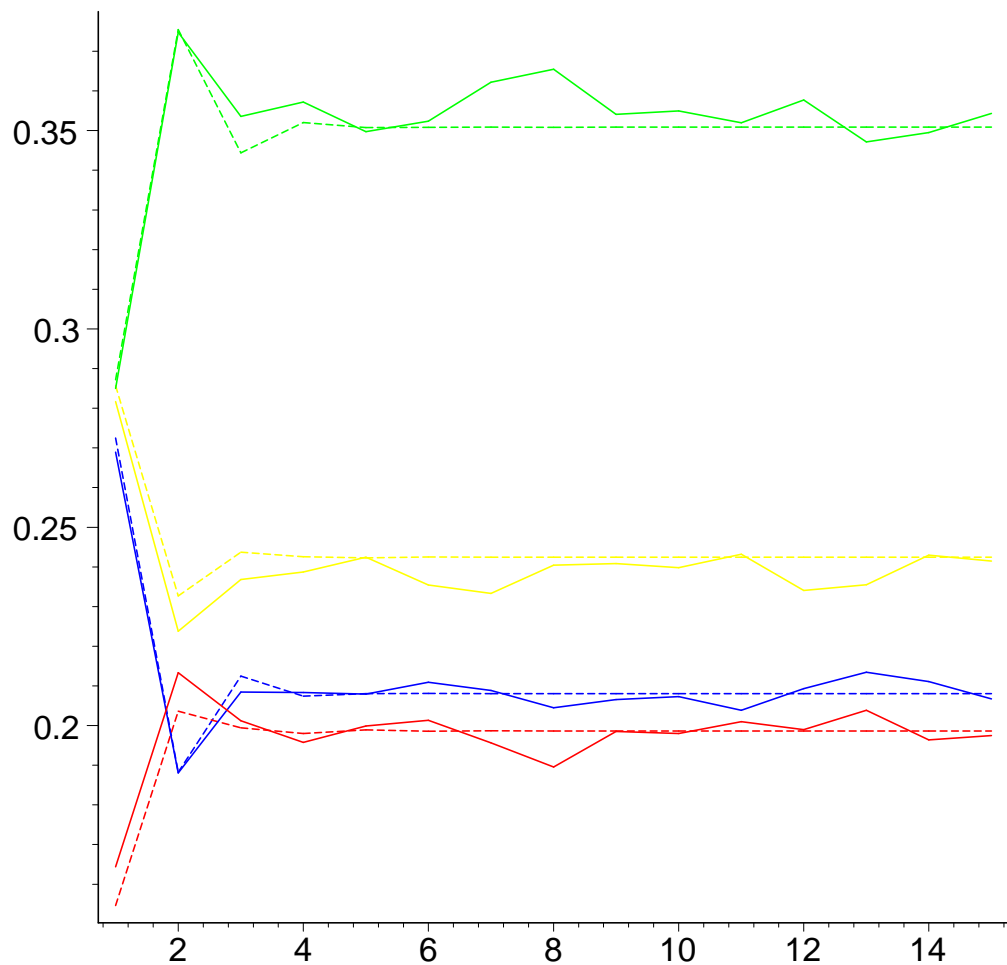
– représentation graphique des résultats

– visualisation de la convergence de la loi simulée vers la loi calculée

```

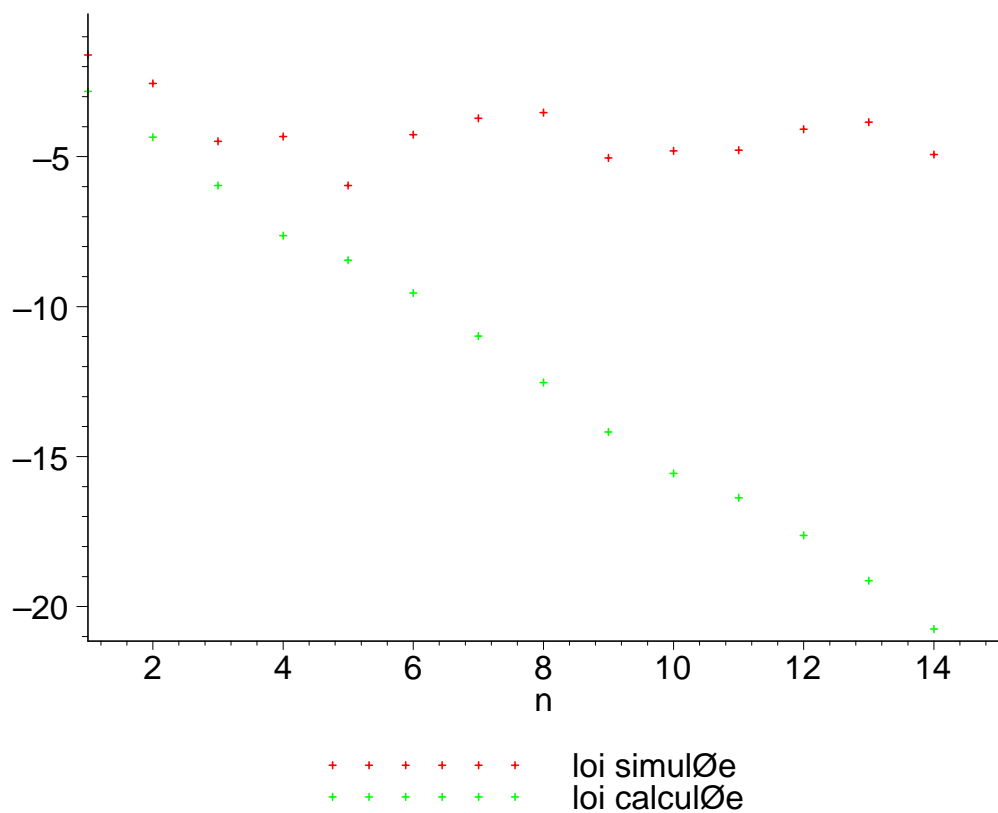
> l1 := [seq( [j-1, mu[1, j]], j=2..t_max+1)]:
l2 := [seq( [j-1, mu[2, j]], j=2..t_max+1)]:
l3 := [seq( [j-1, mu[3, j]], j=2..t_max+1)]:
l4 := [seq( [j-1, mu[4, j]], j=2..t_max+1)]:
p1 := plot({l1, l2, l3, l4},
color=[red,blue,green,yellow]):
l5 := [seq( [j, eta[j][1]], j=1..t_max)]:
l6 := [seq( [j, eta[j][2]], j=1..t_max)]:
l7 := [seq( [j, eta[j][3]], j=1..t_max)]:
l8 := [seq( [j, eta[j][4]], j=1..t_max)]:
p2 := plot({l5, l6, l7, l8},
color=[red,blue,green,yellow], linestyle=DASH):
display({p1,p2});

```



— visualisation de la convergence de ces deux lois vers la loi invariante

```
> f := i -> log(abs(pi[1]-mu[1,i]) + abs(pi[2]-mu[2,i]) +  
  abs(pi[3]-mu[3,i]) + abs(pi[4]-mu[4,i]))):  
g := i -> log(abs(pi[1]-eta[i][1]) + abs(pi[2]-eta[i][2])  
  + abs(pi[3]-eta[i][3]) + abs(pi[4]-eta[i][4]))):  
lf := [seq([n-1,f(n)],n=1..t_max)]:  
lg := [seq([n-1,g(n)],n=1..t_max)]:  
plot([lf, lg],n=1..t_max, style=point, legend=["loi  
simulée", "loi calculée"]);
```



regression linéaire - calcul de l'équation de la droite

```

> a := 'a':
u := 'u':
a := evalf([seq(g(n), n=1..t_max)]):
b := [seq(n, n=1..t_max)]:
res := fit[leastsquare][[n,u]]([b,a]):
assign(res):
pente := diff(u, n);
ordonnee_0 := subs(n=0, u);

pente := -1.251218749
ordonnee_0 := -0.9000436200

```

comparaison entre la pente de la droite et $\log(1 - \inf P)$

```

> c := 1-min(seq(seq(Q[i,j], j = 1 .. 4), i = 1 .. 4)):
evalf(exp(pente)) < evalf(c);

0.2861558321 < 0.9871937405

```

L L

– Spectre de P

```
[ > evalf(Eigenvalues(Q));
```

$$\begin{bmatrix} 1. \\ -0.09904610971 \\ -0.2197850785 + 0.1390958000 I \\ -0.2197850785 - 0.1390958000 I \end{bmatrix}$$

– module des valeurs propres

```
[ > evalf(Map(abs, Eigenvalues(Q)));
```

$$\begin{bmatrix} 1. \\ 0.09904610971 \\ 0.2601021382 \\ 0.2601021382 \end{bmatrix}$$

```
[ > exp(pente);
```

$$0.2563816839$$

```
[ >
```