

## Notes 2, April, Sta624

Example of MC for continuous state space (cont. random variables).  
But still discrete time.

1. Transition probability matrix needs to be replaced by a transition probability kernel:  $p(x|y)$ ; which is a density for any given  $y$  value.

2. in the following example  $p(x|y)$  is taken to be Uniform  $(1-y, 1)$  State space is the interval  $(0, 1)$ .

Example of a MC

$X_0 \sim$  any number between 0 and 1 (or could be from any distribution on  $(0,1)$  interval)

$X_1 \sim \text{Unif}(1 - X_0, 1)$

$X_2 \sim \text{Unif}(1 - X_1, 1)$

.....

$X_n \sim \text{Unif}(1 - X_{n-1}, 1); \dots$

This is a MC.

If the distribution of  $X_n$  is convergent at all (here it does), the stationary distribution must satisfy the following (integral) equation

$$\int p(x|y)f(y)dy = f(x)$$

We may check that distribution  $f(x) = 2x$  solves the above equation.

Therefore  $X_n$ , as  $n$  large, will have a distribution approximately equal to  $f(x) = 2x$  for  $0 < x < 1$ .

Remark: this is the cont. version of the equation  $\pi_0 P = \pi_0$  for discrete MC.

Since

$$p(x|y) = \frac{I_{[1-y < x < 1]}}{y}$$

We compute, using  $f(y) = 2y$ ,

$$\int_0^1 p(x|y)f(y)dy = \int I_{[1-y < x < 1]} 2dy = 2 \int_{1-x}^1 dy = 2x .$$

Notice here we only used the random variables from uniform distributions and end up with a random variable with density  $f(x) = 2x$ .

The idea of MCMC is to use an easy  $p(x|y)$  in the iteration but end up with a random variable with (complicated)  $f(x)$  that we desire.

Suppose

$X_0 \sim$  any number between 0 and 1

$X_1 \sim \text{Unif}(0, 1 - X_0)$

$X_2 \sim \text{Unif}(0, 1 - X_1)$

.....

$X_n \sim \text{Unif}(0, 1 - X_{n-1}),$

Find the stationary distribution  $f(x)$  of this MC. This will also be the approx distribution of  $X_n$  for large  $n$ .