

Reading: Finish reading selected sections of CTCN, and the lecture notes, *Representation theory & Markov Chains and Stochastic Stability*.

Exercises:

- 3 Consider the linear state space model on \mathbb{R}^2 ,

$$X(t+1) = AX(t) + \mathcal{E}(t+1)$$

where the eigenvalues of A are strictly less than one. It is assumed that \mathcal{E} is i.i.d., with zero-mean and finite variance. A course on linear systems theory will tell you that for any matrix $Q > 0$, there is a solution to the discrete-time *Lyapunov equation* $A^T M A = M - Q$, where the 2×2 matrix M is positive definite.

- (i) Obtain a solution to Poisson's equation with $c(x) = x^T Q x$.
 (ii) Obtain a solution to Foster's criterion using the Lyapunov function $V(x) = a \log(1 + h(x))$, where $a > 0$ is a constant, and h is your solution to (i).

- 4 Suppose that R is a *non-negative* $N \times N$ matrix: This means that $R_{ij} \geq 0$ for each $1 \leq i \leq N$. Suppose that s is a column vector, ν is a row vector, satisfying $\nu s > 0$, and $R \geq s \otimes \nu$. That is, $R_{ij} \geq s_i \nu_j$ for each i, j . Finally, suppose that $r > 0$ is a solution to the equation,

$$\nu G s = 1 \quad \text{where} \quad G = \sum_{k=0}^{\infty} r^{-k-1} (R - s \otimes \nu)^k \quad (1)$$

Using the ideas of the lecture on January 27, show that $\mu = \nu G$ and $h = G s$ are left and right eigenvectors of R :

$$\mu R = r \mu, \quad R h = r h$$

Notes: This construction is part of the celebrated theory of Perron & Frobenius — begun approximately one hundred years ago. The scalar r is the *Perron-Frobenius eigenvalue*.

A typical systems application uses $R_{ij} = e^{f_i} P_{ij}$, where P is a transition matrix, and f is envisioned as a function on the state space. The eigenvalue r is a value of the moment generating function found in risk-sensitive control and in large deviations theory for Markov chains. *Why?* Because in this case,

$$R_{ij}^n = \mathbb{E} \left[\exp \left(\sum_{t=0}^{n-1} f(X(t)) \right) \mathbf{1}(X(n) = j) \mid X(0) = i \right]$$

- 5 Suppose that there exists a function s and a probability distribution ν on \mathbf{X} satisfying $P = s \otimes \nu$ (that is, $P(x, y) = s(x)\nu(y)$ for each $x, y \in \mathbf{X}$.) Show that $s \equiv 1$, ν is an invariant measure, and hence \mathbf{X} is i.i.d. with marginal distribution ν .

One more ...

- 6 Suppose that the minorization condition holds, $P \geq s \otimes \nu$, with $s(x) \equiv \epsilon > 0$ constant. This is a form of *Doebelin's condition*¹. Observe that $Q(x, y) := [P(x, y) - \epsilon\nu(y)]/(1 - \epsilon)$ is a transition matrix (let's stick to a countable state space for simplicity). Doebelin's condition can be equivalently expressed $P = \epsilon 1 \otimes \nu + (1 - \epsilon)Q$, or

$$P(x, y) = \epsilon\nu(y) + (1 - \epsilon)Q(x, y), \quad x, y \in \mathsf{X}$$

Show that $G(x, \mathsf{X}) \leq \epsilon^{-1}$ for all x , where the potential kernel is given by,

$$G = \sum_{k=0}^{\infty} (P - s \otimes \nu)^k.$$

Is there an invariant probability measure?

¹Wolfgang Doebelin was a *remarkable person*. Just 10 years ago, his sealed manuscript was opened which revealed that he developed the theory of stochastic differential equations while fighting the Germans in WW2! His formulation predates Ito, and is also simpler.