

Comp Econ: Problem Set 9

Analytical Part

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Exercise 1

I will first show that $\|\mathbf{A}\| \geq \rho(\mathbf{A}'\mathbf{A})$. There exists an $x \in \mathbb{R}^N$ such that $\mathbf{A}'\mathbf{A}x = \rho(\mathbf{A}'\mathbf{A})x$. Then:

$$\begin{aligned}\rho(\mathbf{A}'\mathbf{A})\|x\| &= \langle \rho(\mathbf{A}'\mathbf{A})x, x \rangle \\ &= \langle \mathbf{A}'\mathbf{A}x, x \rangle \\ &= \langle \mathbf{A}x, \mathbf{A}x \rangle \\ &= \|\mathbf{A}x\| \\ &\leq \|\mathbf{A}\|\|x\|\end{aligned}$$

Now I will show the other direction, $\|\mathbf{A}\| \leq \rho(\mathbf{A}'\mathbf{A})$:

$$\begin{aligned}\|\mathbf{A}\|\|x\| &= \|\mathbf{A}x\| = \langle \mathbf{A}x, \mathbf{A}x \rangle = \langle \mathbf{A}'\mathbf{A}x, x \rangle \\ &= \left\langle \sum_i \lambda_i P_i x, x \right\rangle = \sum_i \lambda_i \langle P_i x, x \rangle = \sum_i \lambda_i \langle P_i' P_i x, x \rangle \\ &= \sum_i \lambda_i \langle P_i x, P_i x \rangle = \sum_i \lambda_i \|P_i x\|^2 \\ &\leq \sum_i \rho(\mathbf{A}'\mathbf{A}) \|P_i x\|^2 \\ &= \rho(\mathbf{A}'\mathbf{A}) \sum_i \|P_i x\|^2 \\ &= \rho(\mathbf{A}'\mathbf{A}) \left(\sum_i P_i x x' \right) \\ &= \rho(\mathbf{A}'\mathbf{A}) \|x\|^2\end{aligned}$$

Exercise 2

From Gelfand's formula, we can say that for any $\epsilon > 0$, there exists a $K \in \mathbb{N}$ such that for any $k \geq K$, we have $|||A^k|||^{\frac{1}{k}} - \rho(A) < \epsilon$. Then, for any $k \geq K$:

$$\begin{aligned} \|A^k\|^{1/k} - \rho(A) &< \epsilon \\ \|A^k\|^{1/k} &< \rho(A) + \epsilon \\ \|A^k\| &< (\rho(A) + \epsilon)^k \end{aligned}$$

Now, we can let $r = \rho(A) + \epsilon < 1$. Then, $\|A^k\| < r^k C$ where $C = 1$. Observe, using the properties of geometric sums and the fact that $\|A^k\|$ is finite:

$$\sum_{k=0}^{\infty} \|A^k\| = \sum_{k=0}^K \|A^k\| + \sum_{k=K+1}^{\infty} \|A^k\| < \sum_{k=0}^K \|A^k\| + \sum_{k=K+1}^{\infty} r^k = \sum_{k=0}^K \|A^k\| + \frac{r^{K+1}}{1-r} < \infty$$

Exercise 3

If A is diagonalizable, then it can be written as $A = PVP^{-1}$ where P is a matrix whose columns are the eigenvectors of A and form an orthonormal basis of \mathbb{R}^N , and V is a diagonalizable matrix whose diagonal elements are equal to the eigenvalues of A . Then:

$$\|A^k\| = \|(PVP^{-1})^k\| = \|PV^kP^{-1}\| \leq \|P\| \|V^k\| \|P^{-1}\| \leq \|P\| \|P^{-1}\| \|V\|^k$$

We can also see that:

$$\|V\| = \max_{s.t. \|x\|=1} \|Vx\| = \max_{s.t. \|x\|=1} \sqrt{(\lambda_1 x_1)^2 + \dots + (\lambda_n x_n)^2} = \max_i \sqrt{\lambda_i^2} = \max_i |\lambda_i| = \rho(A)$$

Together, these imply:

$$\|A^k\| \leq \|P\| \|P^{-1}\| \|V\|^k = \|P\| \|P^{-1}\| \rho(A)^k$$

Since $\rho(A) < 1$, we can see that $\|A^k\| \rightarrow 0$ as $k \rightarrow \infty$.

Exercise 4

I will show that for any set of non-negative definite matrices, the limit point is also a non-negative definite matrix.

- Let M_k be a sequence of non-negative definite matrixes such that $M_k \rightarrow M$ and $k \rightarrow \infty$ using the matrix norm, $\|\cdot\|$, and for a contradiction, suppose that the limit point M is non-negative definite.
- We can say that $\|M_k - M\| < \epsilon \ \forall k \geq K \in \mathbb{N}$. Moreover, we can say that for all $\epsilon > 0$ and all $k \geq K$, there exists a $\delta > 0$ such that $\|M_k - M\| < \delta$ implies that $|z'M_k z - z'M z| < \epsilon$.

- Let $\varepsilon = \frac{z'Mz}{2}$. Then $|z'M_kz - z'Mz| < \frac{z'Mz}{2}$, which implies:

$$\begin{aligned} -\frac{z'Mz}{2} &< z'M_kz - z'Mz < \frac{z'Mz}{2} \\ \frac{z'Mz}{2} &< z'M_kz < \frac{3z'Mz}{2} \end{aligned}$$

- But since we assumed $z'Mz < 0$, then $z'M_kz < 0$, which is a contradiction.
- Then the limit point must also be a non-negative definite matrix. This means that all the sequences in the space of non-negative definite matrices must have limit points that fall inside the space. So the space of nonnegative definite matrices is a closed subset of $(\mathcal{M}(n \times n), \|\cdot\|)$.

Exercise 5

Let L be the Lyapunov operator where $LX = AXA' + M$. L is a uniform contraction on $(\mathcal{M}(n \times n), \|\cdot\|)$, therefore it has a unique fixed point, X^* .

1. We know that $X^* = AX^*A' + M$ and we know that M is symmetric. We also know that for any $X^* \in \mathcal{M}(n \times n)$:

$$(X^*)' = (AX^*A' + M)' = (AX^*A')' + M' = A(X^*)'A' + M$$

So we have found another solution to the Lyapunov equation, $(X^*)'$, but we know that there is only a unique solution, therefore $X^* = (X^*)'$. Therefore, if M is symmetric, then X^* is also symmetric.

2. Let the space of nonnegative definite matrices be $\mathcal{ND}(n \times n)$. We proved in Exercise 4 that a nonnegative definite matrix B has the property that $z'Bz \geq 0$ for all $z \in \mathbb{R}^n/\{0\}$. Let M be a nonnegative definite matrix and consider the Lyapunov operator. For any $X \in \mathcal{M}(n \times n)$, AXA' is nonnegative definite. We can also say that $AXA' + M$ is nonnegative definite. Thus, if X is nonnegative definite, the Lyapunov operator is such that: $L : \mathcal{ND}(n \times n) \rightarrow \mathcal{ND}(n \times n)$. From Exercise 4, $\mathcal{ND}(n \times n)$ is a closed subset of $\mathcal{M}(n \times n)$. Applying the corollary to the contraction mapping theorem, we see that X^* , is contained in $\mathcal{ND}(n \times n)$. Therefore, if M is nonnegative definite then X^* is nonnegative definite.
3. Let the space of positive definite matrices be $\mathcal{PD}(n \times n)$ and let M be a positive definite matrix. Consider the Lyapunov operator. Note that for any $X \in \mathcal{M}(n \times n)$, AXA' is nonnegative definite.

Now, notice that $z'(AXA' + M)z = z'AXA'z + z'Mz$. We know that AXA' is nonnegative definite, so $z'AXA'z \geq 0 \implies z'AXA'z + z'Mz > 0$. Thus, we know that the Lyapunov operator takes the space of nonnegative definite matrices to the space of positive definite matrices $L : \mathcal{ND}(n \times n) \rightarrow \mathcal{PD}(n \times n)$.

From above, we know that $\mathcal{ND}(n \times n) \overset{\text{closed}}{\subset} \mathcal{M}(n \times n)$ and $\mathcal{PD}(n \times n) \subset \mathcal{ND}(n \times n)$. This is because the set of positive definite matrices is just the set of nonnegative definite matrices minus all of the matrices that are indefinite i.e. $z'Mz = 0$ for some $z \in \mathbb{R}^n/\{0\}$.

Applying the corollary to the contraction mapping theorem, $L(\mathcal{ND}(n \times n)) \subseteq \mathcal{PD}(n \times n) \subset \mathcal{ND}(n \times n)$ implies that the fixed point, X^* , is contained in $\mathcal{PD}(n \times n)$. Thus, M positive definite $\implies X^*$ positive definite.

Exercise 6

See my Jupyter notebook. In case the plot doesn't show up in the automatically generated .pdf, here it is.

