EE363 Winter 2008-09

Review Session 3

- Generating colored Gaussian variables
- Solving discrete-time Lyapunov equations

Generating colored Gaussian variables

- ullet Suppose we want a random variable $y \sim \mathcal{N}(\bar{y}, \Sigma_y)$
- ullet Have random number generator for $x \sim \mathcal{N}(0, I)$
- \bullet Perform linear transformation y = Ax + b, get $y \sim \mathcal{N}(b, AA^T)$

Generating Gaussian variables with specified covariance

• Can set $A = \Sigma_y^{1/2}$ (symmetric square root of Σ_y), so that

$$\Sigma_{y} = \mathbf{E}(y - \bar{y})(y - \bar{y})^{T}$$

$$= \mathbf{E}(\Sigma_{y}^{1/2}x + \bar{y} - \bar{y})(\Sigma_{y}^{1/2}x - \bar{y} - \bar{y})^{T}$$

$$= \Sigma_{y}^{1/2}\mathbf{E}(xx^{T})\Sigma_{y}^{1/2}$$

$$= \Sigma_{y}$$

- ullet Can also set $A=\Sigma_y^{1/2}Q$, where Q is any orthogonal matrix
- Since $QQ^T=I$, we get $\Sigma_y=\Sigma_y^{1/2}Q\,\mathbf{E}(xx^T)Q^T\Sigma_y^{1/2}=\Sigma_y^{1/2}QQ^T\Sigma_y^{1/2}=\Sigma_y$

Using the Cholesky factorization

- Instead of computing the matrix square root, we can use the Cholesky factorization
- A little cheaper: about a tenth of the computational cost
- With $\Sigma_y = LL^T$, set $y = Lx + \bar{y}$, and

$$\mathbf{E}(y-\bar{y})(y-\bar{y})^T = L\,\mathbf{E}(xx^T)L^T = LL^T = \Sigma_y$$

ullet Can find an orthogonal Q that relates L and $\Sigma_y^{1/2}$, because

$$\Sigma_y = (\Sigma_y^{1/2} Q)(Q^T \Sigma_y^{1/2}) = LL^T$$

so if
$$Q = A^{-1/2}L$$
, then $L = \Sigma_y^{1/2}Q$

- \bullet Therefore, using L is equivalent to orthogonal transformation of x then multiplication by $A^{1/2}$
- Orthogonal transformation of random variable retains second-order statistics
- In Matlab, sqrtm(Sigma_y) or chol(Sigma_y)

Solving Lyapunov equations efficiently

- Used in (upcoming) lecture 13, Linear quadratic Lyapunov theory
- Continuous time, $A^TP + PA + Q = 0$
- Discrete time, $A^T P A P + Q = 0$
- Solving for P with naïve method is $\mathcal{O}(n^6)$
- Using fast method, can solve in $\mathcal{O}(n^3)$
- Will show fast method for solving Sylvester equations (includes Lyapunov equations as special case)

Discrete-time Sylvester operator

• The discrete-time Sylvester operator $S: \mathbf{R}^{n \times n} \to \mathbf{R}^{n \times n}$ is defined as

$$S(X) = AXB - X$$

where A, B, $X \in \mathbf{R}^{n \times n}$.

ullet We will show that the (n^2) eigenvalues of the Sylvester operator are

$$\lambda_i \mu_j - 1, \qquad i, j = 1, \dots, n$$

where $\lambda_1, \ldots, \lambda_n$ are the eigenvalues of A, and μ_1, \ldots, μ_n are the eigenvalues of B.

Eigenvalues of the discrete-time Sylvester operator

• Let v_i be the right eigenvector of A associated with the eigenvalue λ_i . Then we know that

$$Av_i = \lambda_i v_i$$

• Let w_j be the left eigenvector of B associated with the eigenvalue μ_j . Then we know that

$$w_j^T B = \mu_j w_j^T$$

• Let $X = v_i w_j^T$. We will show that X is a (matrix) eigenvector of S associated with the eigenvalue $\lambda_i \mu_j - 1$.

Eigenvalues of the discrete-time Sylvester operator

$$AXB - X = A (u_i w_j^T) B - u_i w_j^T$$

$$= (Au_i) (w_j^T B) - u_i w_j^T$$

$$= (\lambda_i u_i) (\mu_j w_j^T) - u_i w_j^T$$

$$= \lambda_i \mu_j u_i w_j^T - u_i w_j^T$$

$$= (\lambda_i \mu_j - 1) u_i w_j^T$$

$$= (\lambda_i \mu_j - 1) X$$

Then we have shown that

$$S(X) = (\lambda_i \mu_j - 1) X$$

which means that X is an eigenvector of S associated with the eigenvalue $(\lambda_i \mu_i - 1)$.

Discrete-time Sylvester operator (cnt'd)

- S is singular if and only if there exists a nonzero X with S(X)=0.
- S is nonsingular if and only if, for all i, j, $\lambda_i \mu_j \neq 1$.
- ullet If A and B are stable then S is nonsingular.

Discrete-time Sylvester equation

• The discrete-time Sylvester equation is

$$AXB - X = C$$

where A, B, C, $X \in \mathbf{R}^{n \times n}$.

- The Sylvester equation has a unique solution for any C if and only if S is non-singular, which occurs if and only if $\lambda_i \mu_j \neq 1$ for all i, $j = 1, \ldots, n$.
- The Sylvester equation can be rewritten as a set of n^2 equations in n^2 variables $\mathcal{A}vec(X) = vec(C)$, which can be solved in $O(n^6)$ operations.

Solving the discrete-time Sylvester equation

Suppose that A and B are diagonalizable. Then plugging in $A = T\Lambda T^{-1}$ and $B = SMS^{-1}$ where $\Lambda = \mathbf{diag}(\lambda_1, \dots, \lambda_n)$ and $M = \mathbf{diag}(\mu_1, \dots, \mu_n)$, we get

$$(T\Lambda T^{-1})X(SMS^{-1}) - X = C.$$

Multiplying on the left by T^{-1} and on the right by S, we obtain

$$\Lambda(T^{-1}XS)M - (T^{-1}XS) = T^{-1}CS.$$

Let $\tilde{X} = T^{-1}XS$ and $\tilde{C} = T^{-1}CS$. Then we have

$$\Lambda \tilde{X}M - \tilde{X} = \tilde{C}.$$

Solving the discrete-time Sylvester equation (cnt'd)

Denoting the (i,j)th entry of \tilde{C} as \tilde{c}_{ij} and of \tilde{X} as \tilde{x}_{ij} , the equation reads as

$$\lambda_i \mu_j \tilde{x}_{ij} - \tilde{x}_{ij} = \tilde{c}_{ij},$$

which means that

$$\tilde{x}_{ij} = \frac{\tilde{c}_{ij}}{\lambda_i \mu_j - 1}.$$

Finally, once we compute \tilde{X} , we take $X = T\tilde{X}S^{-1}$.

Note that by exploiting the structure available in the Sylvester operator, we were able to solve the discrete-time Sylvester equation in $O(n^3)$ operations (since the eigenvectors and eigenvalues of a diagonalizable matrix can be found in $O(n^3)$ operations.)

Discrete-time Lyapunov operator

The discrete-time Lyapunov operator is a special case of the discrete-time Sylvester operator:

$$\mathcal{L}(P) = A^T P A - P.$$

ullet L is nonsingular if and only if

$$\lambda_i \lambda_j \neq 1$$
 $i, j = 1, \dots, n,$

where $\lambda_1, \ldots, \lambda_n$ are the eigenvalues of A.

• If A is stable then \mathcal{L} is nonsingular.

Discrete-time Lyapunov equation

The discrete-time Lyapunov equation is

$$A^T P A - P + Q = 0.$$

- ullet If A is stable then for any Q there is exactly one solution P to the discrete-time Lyapunov equation.
- ullet If A is diagonalizable, i.e. if $A=T\Lambda T^{-1}$, then $P=T\tilde{P}T^{-1}$, where

$$\tilde{p}_{ij} = \frac{-\tilde{q}_{ij}}{\lambda_i \lambda_j - 1},$$

with
$$\tilde{Q} = T^{-1}QT$$
.

• The solution can be found efficiently in $O(n^3)$ operations.

Continuous-time Sylvester operator

The continuous-time Sylvester operator $S: \mathbf{R}^{n \times n} \to \mathbf{R}^{n \times n}$ is defined as

$$S(X) = AX + XB$$

where A, B, $X \in \mathbf{R}^{n \times n}$. The (n^2) eigenvalues of the Sylvester operator are

$$\lambda_i + \mu_j, \qquad i, j = 1, \dots, n$$

where $\lambda_1, \ldots, \lambda_n$ are the eigenvalues of A, and μ_1, \ldots, μ_n are the eigenvalues of B.

Continuous-time Sylvester operator (cnt'd)

- S is singular if and only if there exists a nonzero X with S(X)=0.
- S is nonsingular if and only if, for all i, j, $\lambda_i \neq -\mu_j$, or equivalently, if and only if A and -B share no eigenvalues.
- ullet If A and B are stable, then S is nonsingular.

Continuous-time Sylvester equation

• The continuous-time Sylvester equation is

$$AX + XB = C$$

where A, B, C, $X \in \mathbf{R}^{n \times n}$.

- The Sylvester equation has a unique solution for any C if and only if S is non-singular, which occurs if and only if for all i, j, $\lambda_i \neq -\mu_j$.
- The Sylvester equation can be rewritten as a set of n^2 equations in n^2 variables $\mathcal{A}vec(X) = vec(C)$, which can be solved in $O(n^6)$ operations.
- If we take advantage of the structure of the Sylvester operations, it can be solved in $O(n^3)$ operations.

Continuous-time Lyapunov operator

The *continuous-time Lyapunov operator* is a special case of the continuous-time Sylvester operator:

$$\mathcal{L}(P) = A^T P + P A.$$

 \mathcal{L} is nonsingular if and only if A and -A have no common eigenvalues.

- If A is stable, then \mathcal{L} is nonsingular.
- ullet If A has an imaginary eigenvalue, then ${\mathcal L}$ is singular.

Continuous-time Lyapunov equation

The continuous-time Lyapunov equation is

$$A^T P + PA + Q = 0.$$

- ullet If A is stable then for any Q there is exactly one solution P to the continuous-time Lyapunov equation.
- If A is diagonalizable, i.e. if $A=T\Lambda T^{-1}$, then $P=T\tilde{P}T^{-1}$, where

$$\tilde{p}_{ij} = \frac{-\tilde{q}_{ij}}{\lambda_i + \lambda_j},$$

with
$$\tilde{Q} = T^{-1}QT$$
.

• The solution can be found efficiently in $O(n^3)$ operations.

Solving Sylvester equations in MATLAB

- X = dlyap(A,B,-C) solves the discrete-time Sylvester equation AXB X = C.
- P = dlyap(A',Q) solves the discrete-time Lyapunov equation $A^TPA P + Q = 0$.
- X = lyap(A,B,-C) solves the continuous-time Sylvester equation AX + XB = C.
- X = lyap(A',Q) solves the continuous-time Lyapunov equation $A^TP + PA + Q = 0$.