

Announcements

Tuesday, April 10

Exams and assignments

- ▶ *Midterm 3*: Friday 13th, 3:00-3:50
Sections 6.1-6.6 and 7.1-7.2
- ▶ *Optional assignment*: Friday 20th, midnight
Due by email: laura.eslava@math.gatech.edu
- ▶ *Final Exam*: Thursday 26th, 2:50-5:40 (This lecture room)
Comprehensive, all sections covered in class

Section N3

- ▶ Quiz missed will have next quiz's grade copied.
- ▶ *TA for last weeks*: Andrew Fu
- ▶ Office Hours: Skiles 230
Thursday 12th: 1:30-4:00 pm
Thursday 19th: 2:00-4:00 pm

Section 7.3

Constrained Optimization

Motivation: How to allocate resources

Problem: The government wants to repair

- ▶ w_1 hundred miles of public roads
- ▶ w_2 hundred acres of parks

Resources are limited, so cannot work on more than

- ▶ 3 miles of roads or
- ▶ 2 acres of park;
- ▶ *general condition* is:

$$4w_1^2 + 9w_2^2 \leq 36$$

How to allocate resources?

Utility function: Considering overall benefits, want to maximize

$$q(w_1, w_2) = w_1 w_2.$$

(i.e. Do not focus solely on roads nor parks)

How would you *maximize utility* $q(w_1, w_2)$?

Constrained Optimization

Optimization problems

Maximize (or minimize) the value of a given function.

- ▶ This is a broad and important area,
- ▶ here we only focus on *quadratic functions*.

Example

What is the maximum value possible for

- ▶ $Q(x) = 3x_1^2 + 3x_2^2?$
under the constraint $\|x\|^2 = 1$
- ▶ $Q(x) = 3x_1^2 + 7x_2^2?$
under the constraint $\|x\|^2 = 1$

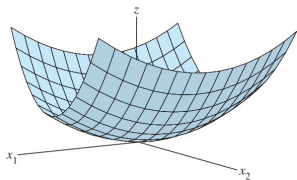


FIGURE 1 $z = 3x_1^2 + 7x_2^2$.

The constraint in these optimization problems

We will keep the restriction that vectors x in \mathbf{R}^n have unit length;

$$\|x\| = 1, \quad x \cdot x = 1 \quad x^T x = 1$$

or more commonly used: $x_1^2 + x_2^2 + \cdots + x_n^2$.

Example

$$Q(x) = 3x_1^2 + 7x_2^2$$

Plot this function in 3-dimension as:

$$\begin{pmatrix} x_1 \\ x_2 \\ Q(x) \end{pmatrix}$$

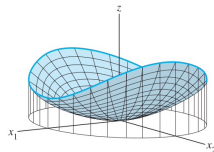


FIGURE 2 The intersection of $z = 3x_1^2 + 7x_2^2$ and the cylinder $x_1^2 + x_2^2 = 1$.

The constrained optimization problem

Given a **quadratic form** $Q(x)$, restricted to *unit vectors*,

What is the maximum and minimum values of $Q(x)$,
which vectors attain such extremes?

An easy case: no cross-product

Example

If $Q(x) = 3x_1^2 + 7x_2^2$, and constraint is $x^T x = 1$.

- ▶ What are the *maximum and minimum* values of $Q(x)$,
- ▶ which *vectors* attain such values?

The constraint means $x_1^2 + x_2^2 = 1$, so for any such vector x :

- ▶ $Q(x) = 3x_1^2 + 7x_2^2 \leq 7x_1^2 + 7x_2^2 \leq 7(x_1^2 + x_2^2) = 7$

attained by vectors $\pm \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

- ▶ $Q(x) = 3x_1^2 + 7x_2^2 \geq 3x_1^2 + 3x_2^2 \geq 3(x_1^2 + x_2^2) = 3$

attained by vectors $\pm \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

Another easy case: no cross-product

Example

If $Q(x) = 2x_1^2 + 4x_2^2 + x_3^2$, and constraint is $x^T x = 1$.

- ▶ What are the *maximum and minimum* values of $Q(x)$,
- ▶ which *vectors* attain such values?

The associated matrix is $A = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ with eigenvalues 1, 2, 4.

- ▶ The maximum value is 4, attained by $\pm \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$
- ▶ The minimum value is 1, attained by $\pm \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$

Eigenvalues

Answer seems: The largest and smallest eigenvalues (and eigenvectors) of A . How does it work in general?

The Constrained Optimization theorem

Theorem

Let A be a symmetric matrix and $Q(x) = x^T A x$ a quadratic function

- ▶ **Maximum:** the maximum value of $Q(x)$ subject to $x^T x = 1$ equals the **largest eigenvalue** M of A .

This maximum is attained by an eigenvector of A corresponding to M .

- ▶ **Minimum:** the minimum value of $Q(x)$ subject to $x^T x = 1$ equals the **smallest eigenvalue** m of A .

This minimum is attained by an eigenvector of A corresponding to m .

How to use this information? To find maximum/minimum values of $Q(x)$, under *restriction* $x^T x = 1$:

- ▶ Find the eigenvalues of A , list them in decreasing order $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$.
- ▶ Then maximum is $M = \lambda_1$ and minimum is $m = \lambda_n$.

Why it works for all quadratic functions?

From Section 7.2

Recall all quadratic functions $Q(x)$ have

- ▶ A symmetric matrix associated A ,
- ▶ An orthogonal diagonalization for $A = PDP^T$,
- ▶ *Form is equivalent to $\widehat{Q}(y) = \lambda_1 y_1^2 + \lambda_2 y_2^2 + \cdots + \lambda_n y_n^2$,*
under a suitable change of variables $x = Py$

The problem for $\widehat{Q}(y)$: Maximum is largest of λ_i 's, say λ_1 . Then vector attaining maximum is e_1 .

The problem for $Q(x)$: can use $\widehat{Q}(y)$ because P orthonormal!

The maximum is $\widehat{Q}(e_1) = Q(Pe_1) = \lambda_1$ attained by Pe_1 (first column of P).

Example

Example

What is the maximum value of $Q(x) = x^T Ax$ subject to $x^T x = 1$,

$$A = \begin{pmatrix} 3 & 2 & 1 \\ 2 & 3 & 1 \\ 1 & 1 & 4 \end{pmatrix}.$$

For maximum value: compute the characteristic equation of A

$$\det(A - \lambda I) = 0 = (\lambda - 6)(\lambda - 3)(\lambda - 1).$$

Then the maximum value is 6.

For unit vector attaining $Q(x) = 6$: Find eigenvector of A corresponding to 6,
and normalize it!

Get both using a decomposition of A ...

Have access to orthogonal diagonalization?

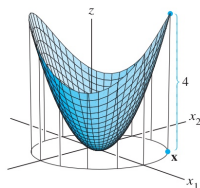
Example

What is the vector attaining the maximum value of $Q(x) = x^T A x$ subject to $x^T x = 1$, $A = \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$.

If you have the *orthogonal diagonalization of A*:

$$\begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix}$$

- ▶ The **maximum** value is 4
- ▶ and the *vectors attaining* such value are $\pm \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$



The maximum value of $Q(x)$ subject to $x^T x = 1$ is 4.

Other eigenvalues/eigenvectors

Let A be a symmetric matrix. Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the eigenvalues of A listed in decreasing order.

Let $A = PDP^T$ be an orthogonal diagonalization where diagonal entries in D are $\lambda_1, \lambda_2, \dots, \lambda_n$ and columns in P are u_1, u_2, \dots, u_n .

- ▶ The maximum value of $x^T Ax$ subject to the constraints

$$x^T x = 1 \quad \text{and} \quad x^T u_1 = 0$$

is the *second largest* eigenvalue λ_2 , attained by $\pm u_2$.

- ▶ The maximum value of $x^T Ax$ subject to the constraints

$$x^T x = 1 \quad \text{and} \quad x^T u_1 = 0, \quad x^T u_2 = 0$$

is the *third largest* eigenvalue λ_3 , attained by $\pm u_3$.

- ▶ Can you see a pattern?

Back to the application: setup problem

Problem: The government wants to repair w_1 hundred miles of public roads and w_2 hundred acres of parks.

1. **Maximize** work done! constrain to $4w_1 + 9w_2 = 36$

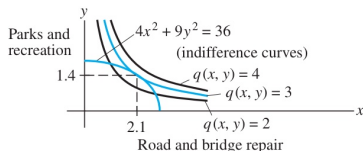
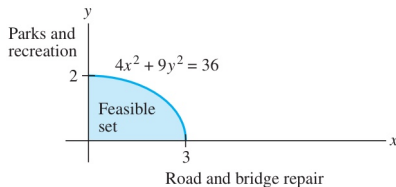


FIGURE 4 The optimum public works schedule is $(2.1, 1.4)$.

2. Fit to quadratic optimization template (additional change of variables)

$$x_1 = \frac{w_1}{3} \quad x_2 = \frac{w_2}{2}, \quad \text{new constraint: } x_1^2 + x_2^2 = 1.$$

3. **New utility function:** subject to $x_1^2 + x_2^2 = 1$, maximize

$$Q(x_1, x_2) = (3x_1)(2x_2) = 6x_1x_2.$$

Back to the application: interpretation

The associated matrix A to $Q(x_1, x_2) = 6x_1x_2$ has orthogonal diagonalization:

$$A = \begin{pmatrix} 0 & 3 \\ 3 & 0 \end{pmatrix} = PDP^T;$$

with

$$P = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}, \quad D = \begin{pmatrix} 3 & 0 \\ 0 & 3 \end{pmatrix}.$$

This means:

- ▶ $Q(x_1, x_2)$ is **maximized when** $x_1 = x_2 = \frac{1}{\sqrt{2}}$,
- ▶ and the utility *function has value 3*.

Translates to:

- ▶ The utility function $q(w_1, w_2)$ is **maximized when** $w_1 = \frac{3}{\sqrt{2}}$ **and** $w_2 = \frac{2}{\sqrt{2}}$,
- ▶ and the utility function *has the same value 3*.