

C&O 355  
Mathematical Programming  
Fall 2010  
Lecture 6

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# Polyhedra

- **Definition:** For any  $a \in \mathbb{R}^n$ ,  $b \in \mathbb{R}$ , define

$$H_{a,b} = \{ x \in \mathbb{R}^n : a^\top x = b \}$$

**Hyperplane**

$$H_{a,b}^+ = \{ x \in \mathbb{R}^n : a^\top x \geq b \}$$

$$H_{a,b}^- = \{ x \in \mathbb{R}^n : a^\top x \leq b \}$$

**Halfspaces**

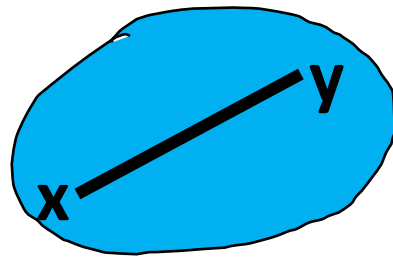
- **Def:** Intersection of finitely many halfspaces is a **polyhedron**
- **Def:** A **bounded** polyhedron is a **polytope**,  
i.e.,  $P \subseteq \{ x : -M \leq x_i \leq M \ \forall i \}$  for some scalar  $M$

- So the feasible region of LP  
is polyhedron  $P = \bigcap_{i=1}^m H_{a_i, b_i}^-$

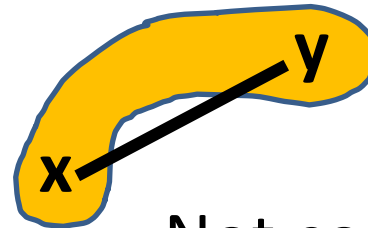
|  |
|--|
| $\begin{array}{ll} \max & c^\top x \\ \text{s.t.} & a_i^\top x \leq b_i \quad \forall i = 1 \dots m \end{array}$ |
|--|

# Convex Sets

- **Def:** Let  $x_1, \dots, x_k \in \mathbb{R}^n$ .  
Let  $\alpha_1, \dots, \alpha_k$  satisfy  $\alpha_i \geq 0$  for all  $i$  and  $\sum_i \alpha_i = 1$ .  
The point  $\sum_i \alpha_i x_i$  is a **convex combination** of the  $x_i$ 's.
- **Def:** A set  $C \subseteq \mathbb{R}^n$  is **convex** if for every  $x, y \in C$  and  $\forall \alpha \in [0, 1]$ , the convex combination  $\alpha x + (1 - \alpha)y$  is in  $C$ .



Convex



Not convex

- **Claim 1:** Any halfspace is convex.
- **Claim 2:** The intersection of any number of convex sets is convex.
- **Corollary:** Every polyhedron is convex.

# Convex Functions

- Let  $C \subseteq \mathbb{R}^n$  be convex.
- **Def:**  $f : C \rightarrow \mathbb{R}$  is a **convex function** if
$$f(\alpha x + (1-\alpha)y) \leq \alpha f(x) + (1-\alpha)f(y) \quad \forall x, y \in C$$
- **Claim:** Let  $f : C \rightarrow \mathbb{R}$  be a convex function, and let  $a \in \mathbb{R}$ . Then
$$\{ x \in C : f(x) \leq a \}$$
 (the “sub-level set”) is convex.
- **Example:** Let  $f(x) = \|x\| = \sqrt{x^T x}$ . Then  $f$  is convex.
- **Corollary:** Let  $B = \{ x : \|x\| \leq 1 \}$ . (The Euclidean Ball)  
Then  $B$  is convex.

# Affine Maps

- **Def:** A map  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is called an **affine map** if  $f(x) = Ax + b$  for some matrix  $A$  and vector  $b$ .
- **Fact:** Let  $C \subseteq \mathbb{R}^n$  have defined volume. Let  $f(x) = Ax + b$ . Then  $\text{vol } f(C) = |\det A| \cdot \text{vol } C$ .
- **Claim 1:** Let  $C \subseteq \mathbb{R}^n$  be convex and let  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be an affine map. Then  $f(C) = \{ f(x) : x \in C \}$  is convex.
- **Question:** If  $P \subseteq \mathbb{R}^n$  is a polyhedron and  $f$  is an affine map, is it true that  $f(P)$  is a polyhedron?
- **Answer:** Yes, but it's not so easy to prove...



[Joseph Fourier](#)

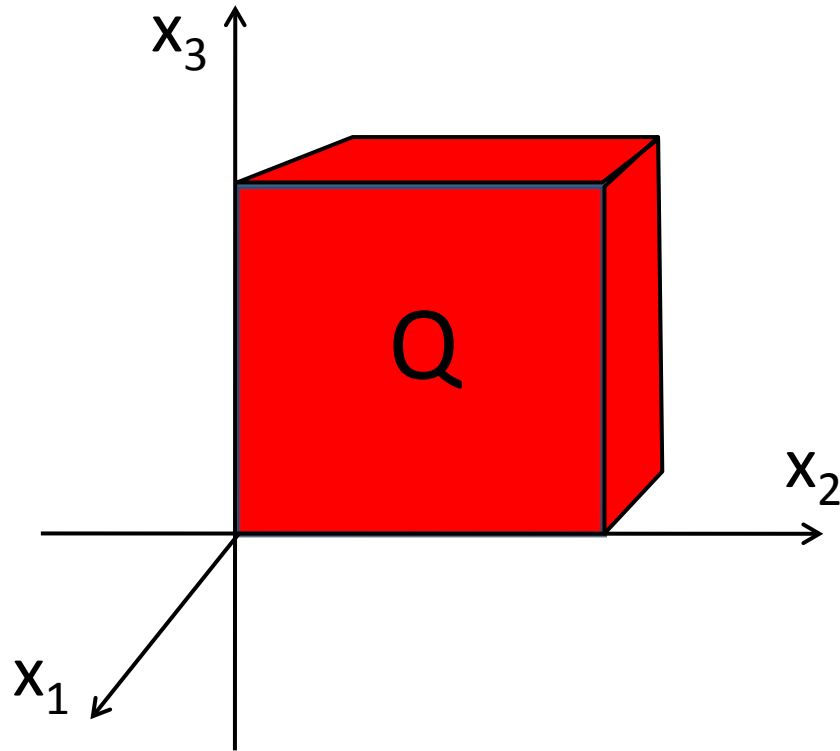
# Fourier-Motzkin Elimination



[Theodore Motzkin](#)

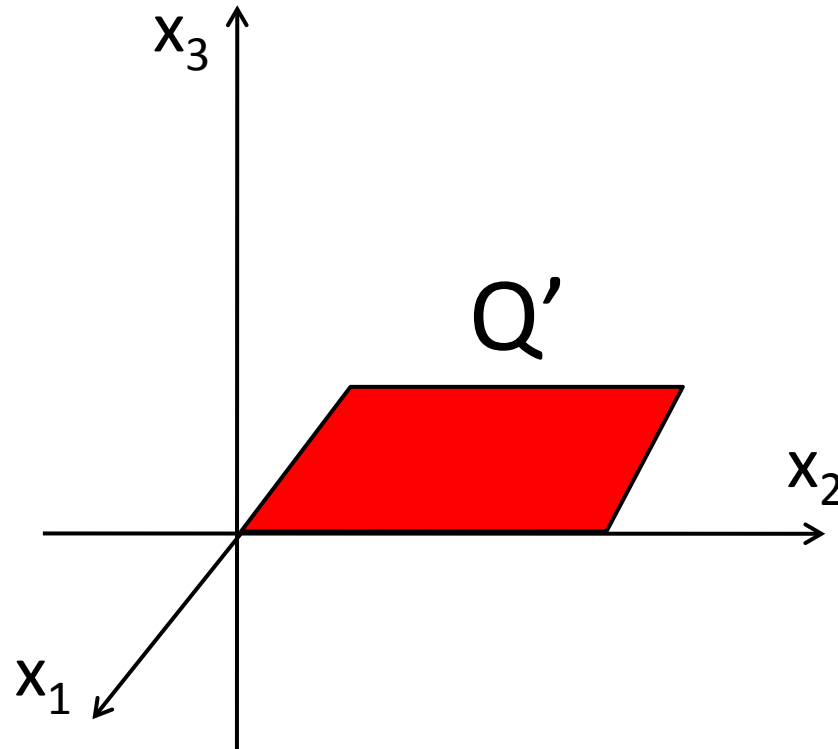
- Given a polyhedron  $Q \subseteq \mathbb{R}^n$ ,  
we want to find the set  $Q' \subseteq \mathbb{R}^{n-1}$  satisfying
$$(x_1, \dots, x_{n-1}) \in Q' \iff \exists x_n \text{ s.t. } (x_1, \dots, x_{n-1}, x_n) \in Q$$
- $Q'$  is called the **projection** of  $Q$  onto first  $n-1$  coordinates
- Fourier-Motzkin Elimination constructs  $Q'$  by generating (finitely many) constraints from the constraints of  $Q$ .
- **Corollary:**  $Q'$  is a polyhedron.

# Elimination Example



- Project  $Q$  onto coordinates  $\{x_1, x_2\}$ ...

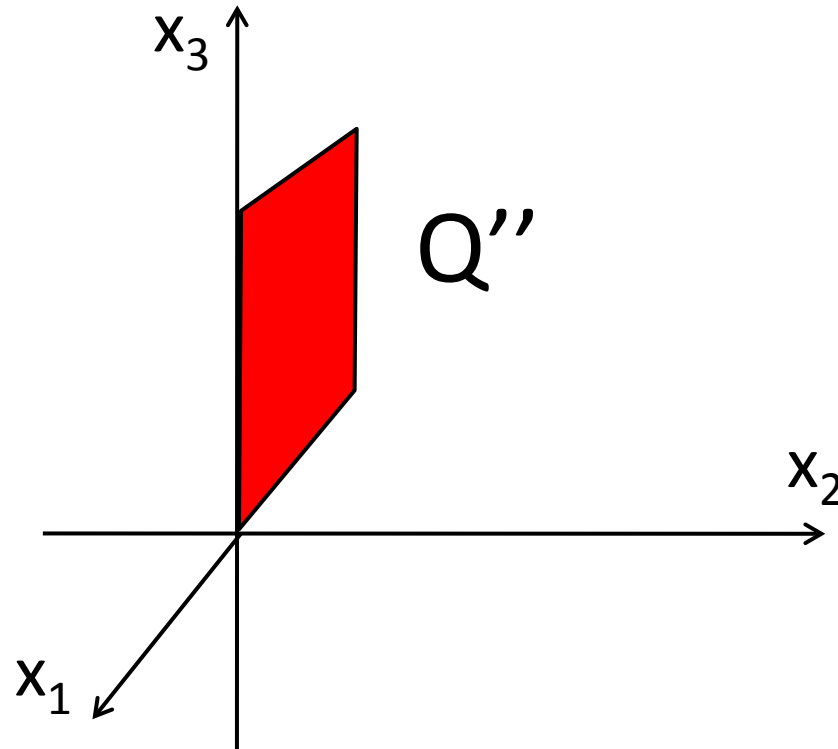
# Elimination Example



- Project  $Q$  onto coordinates  $\{x_1, x_2\}$ ...
- **Fourier-Motzkin:**  $Q'$  is a polyhedron.
- Of course, the ordering of coordinates is irrelevant.

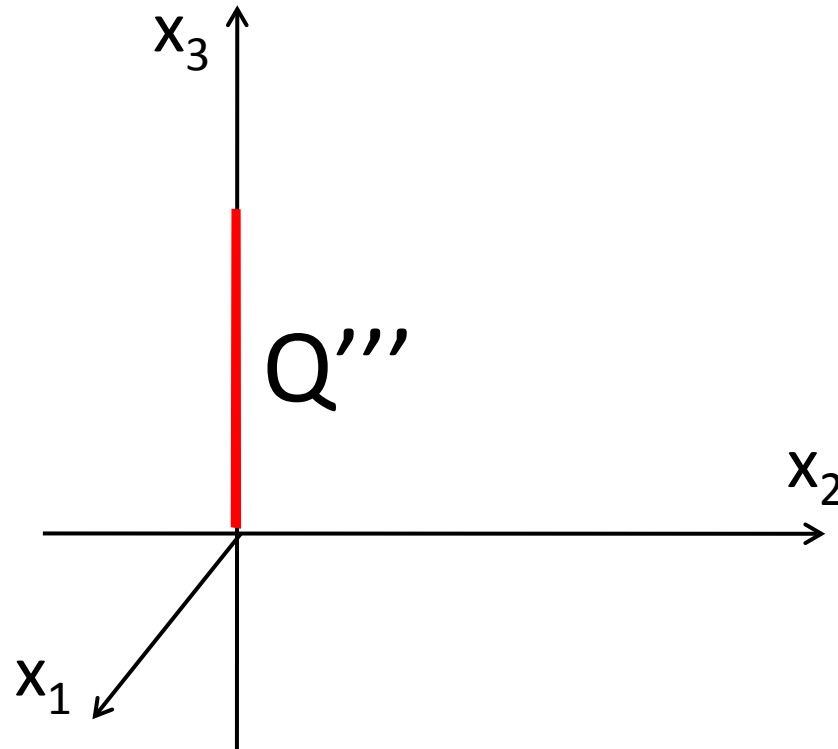


# Elimination Example



- Of course, the ordering of coordinates is irrelevant.
- **Fourier-Motzkin:**  $Q''$  is also a polyhedron.
- I can also apply Elimination twice...

# Elimination Example



- **Fourier-Motzkin:**  $Q'''$  is also a polyhedron.

# Projecting a Polyhedron Onto Some of its Coordinates

- **Lemma:** Given a polyhedron  $Q \subseteq \mathbb{R}^n$ .

Let  $S = \{s_1, \dots, s_k\} \subseteq \{1, \dots, n\}$  be any subset of the coordinates.

Let  $Q_S = \{(\mathbf{x}_{s_1}, \dots, \mathbf{x}_{s_k}) : \mathbf{x} \in Q\} \subseteq \mathbb{R}^k$ .

In other words,  $Q_S$  is projection of  $Q$  onto coordinates in  $S$ .

Then  $Q_S$  is a polyhedron.

- **Proof:**

Direct from Fourier-Motzkin Elimination.

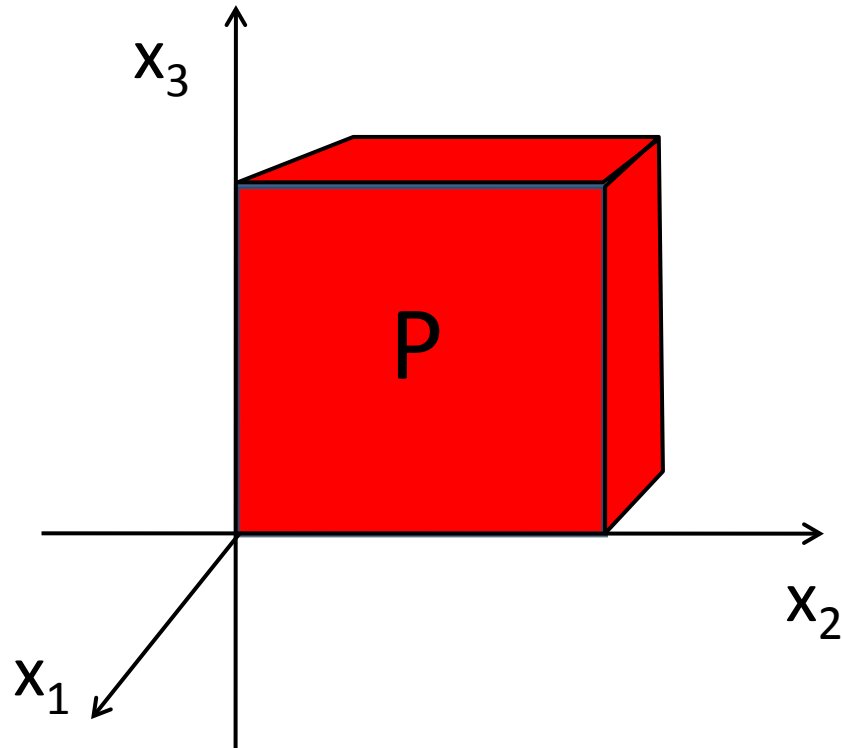
Just eliminate all coordinates not in  $S$ . ■

# Linear Transformations of Polyhedra

- **Lemma:** Let  $P = \{x : Ax \leq b\} \subseteq \mathbb{R}^n$  be a polyhedron.  
Let  $M$  be any matrix of size  $p \times n$ .  
Let  $Q = \{Mx : x \in P\} \subseteq \mathbb{R}^p$ . Then  $Q$  is a polyhedron.

Let  $M =$

|    |   |   |
|----|---|---|
| 1  | 0 | 0 |
| -1 | 1 | 0 |
| 0  | 0 | 1 |



# Linear Transformations of Polyhedra

- **Lemma:** Let  $P = \{x : Ax \leq b\} \subseteq \mathbb{R}^n$  be a polyhedron.

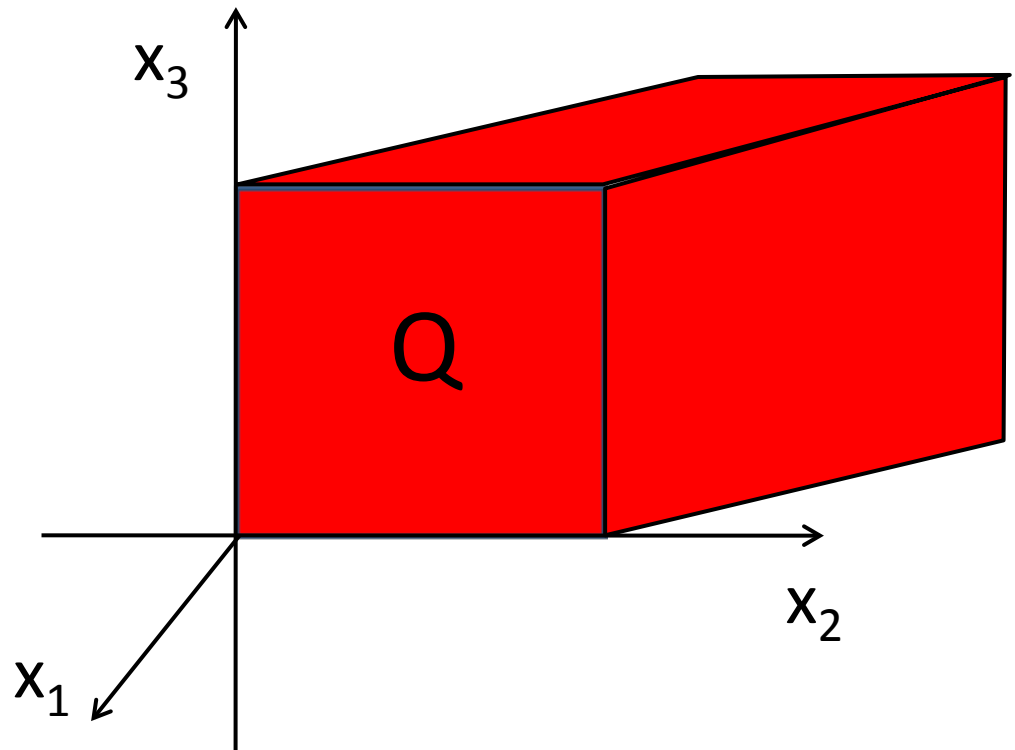
Let  $M$  be any matrix of size  $p \times n$ .

Let  $Q = \{Mx : x \in P\} \subseteq \mathbb{R}^p$ . Then  $Q$  is a polyhedron.

Geometrically obvious, but not easy to prove...

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# Linear Transformations of Polyhedra

- **Lemma:** Let  $P = \{ x : Ax \leq b \} \subseteq \mathbb{R}^n$  be a polyhedron.

Let  $M$  be any matrix of size  $p \times n$ .

Let  $Q = \{ Mx : x \in P \} \subseteq \mathbb{R}^p$ . Then  $Q$  is a polyhedron.

Geometrically obvious, but not easy to prove...

...unless you know Fourier-Motzkin Elimination!

- **Proof:**

Let  $P' = \{ (x, y) : Mx = y, Ax \leq b \} \subseteq \mathbb{R}^{n+p}$ , where  $x \in \mathbb{R}^n$ ,  $y \in \mathbb{R}^p$ .

$P'$  is obviously a polyhedron.

Note that  $Q$  is projection of  $P'$  onto  $y$ -coordinates.

By previous lemma,  $Q$  is a polyhedron. ■

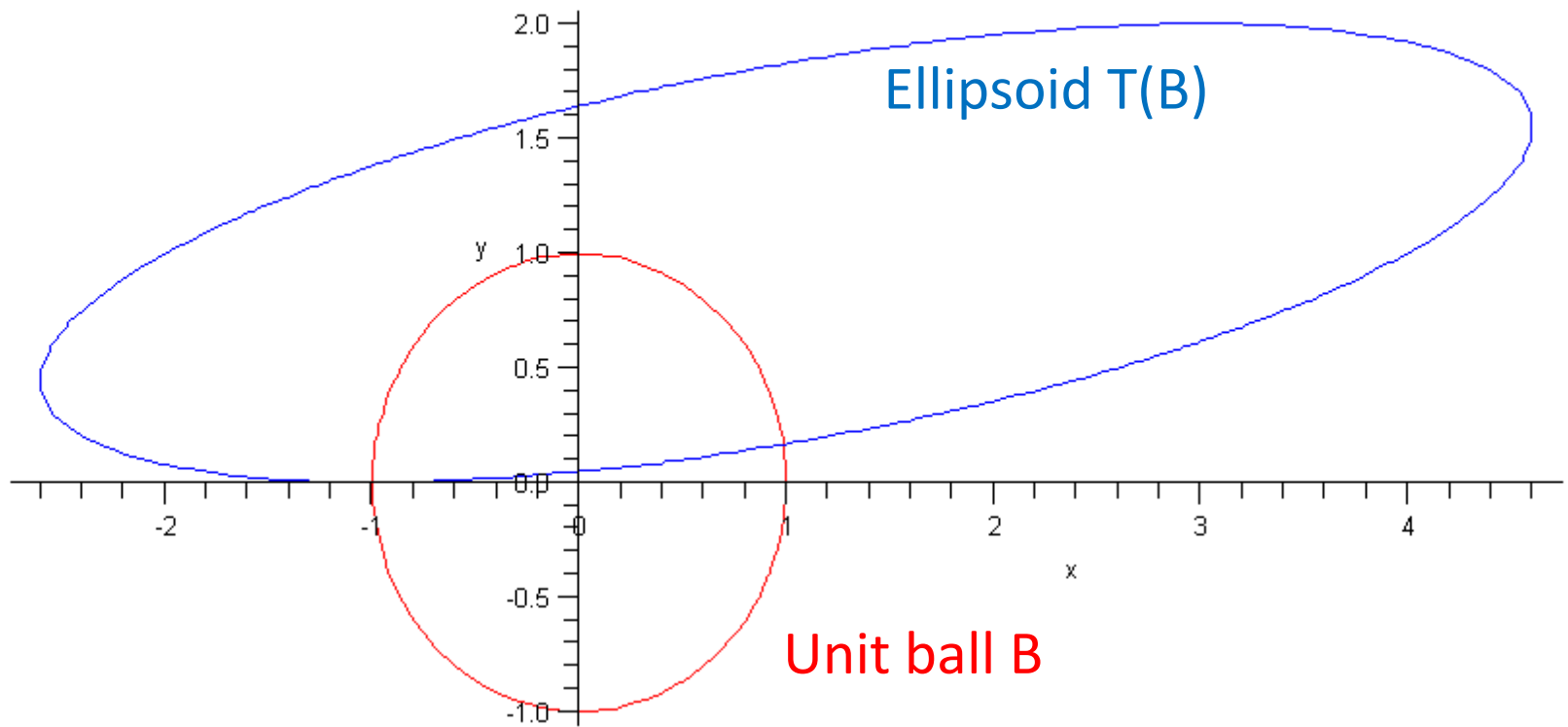
# Ellipsoids

- **Def:** Let  $B = \{ x : \|x\| \leq 1 \}$ . Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be an affine map. Then  $f(B)$  is an **ellipsoid**.
- We restrict to the case  $n=m$  and  $f$  invertible.  
(i.e.,  $f(x) = Ax+b$  where  $A$  is square and non-singular)
- **Claim 2:**  $f(B) = \{ x \in \mathbb{R}^n : (x-b)^T A^{-T} A^{-1} (x-b) \leq 1 \}$ .

# 2D Example

Define  $T(x) = Ax + b$  where  $A = \begin{pmatrix} 3 & 2 \\ 0 & 1 \end{pmatrix}$  and  $b = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

```
implicitplot([x^2+y^2=1, (x-1)^2-4*(x-1)*(y-1)+13*(y-1)^2=9], x=-5..5, y=-5..5,  
numpoints=10000, color=[red,blue] );
```



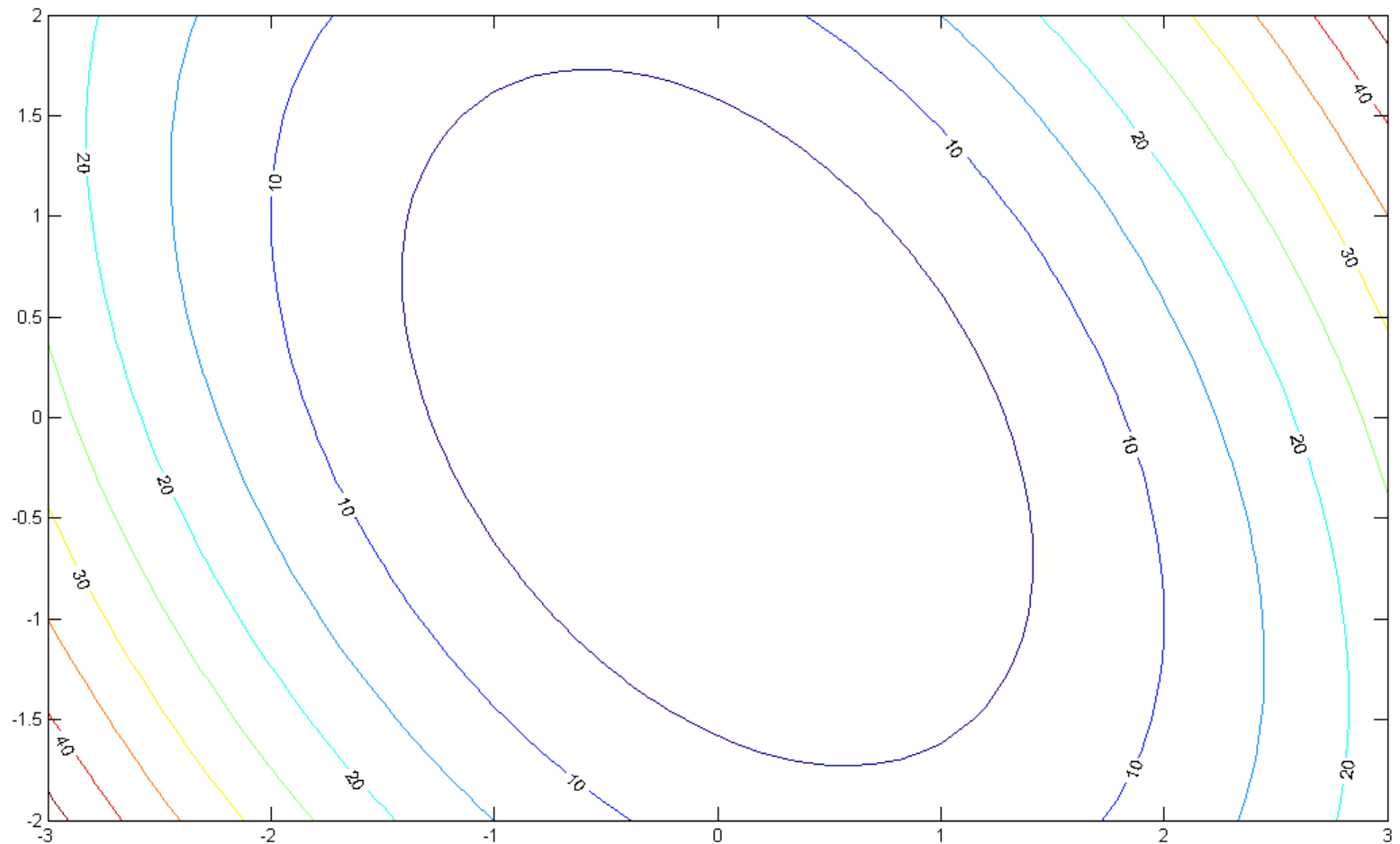


# Positive Semi-Definite Matrices

- **Def:** A real, symmetric matrix  $M$  is called **positive semi-definite** if
$$M = V^T V, \text{ for some matrix } V \text{ (not necessarily square).}$$
If  $M$  is also non-singular it is called **positive definite**.
- From now on “positive definite” means “real, symmetric, positive definite”.
- **Fact:** If  $M$  is positive definite, then  $M^{-1}$  is also positive definite.
- **Fact:** If  $M$  is positive definite, there is a unique positive definite matrix  $M^{1/2}$  such that  $M = M^{1/2} M^{1/2}$ .  $M^{1/2}$  is called the **square root** of  $M$ .
- **Claim:** If  $M$  is positive definite then  $(M^{1/2})^{-1} = (M^{-1})^{1/2}$ .

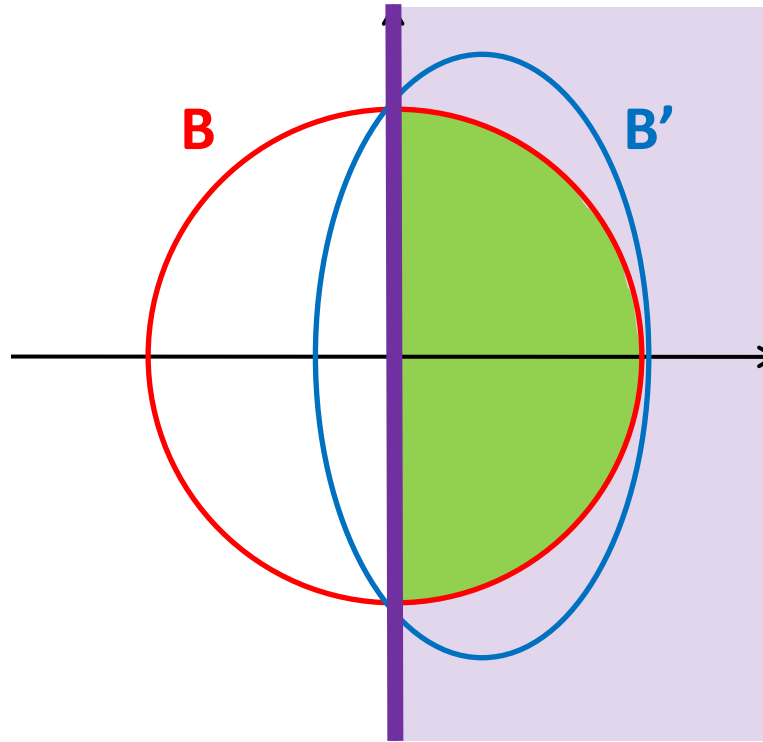
# More on Ellipsoids

- We're studying ellipsoids of the form
$$\{ x \in \mathbb{R}^n : (x-b)^T A^{-T} A^{-1} (x-b) \leq 1 \},$$
for some (non-singular) matrix  $A$  and vector  $b$ .
- Equivalently, this is ellipsoids of the form
$$E(M,b) = \{ x \in \mathbb{R}^n : (x-b)^T M^{-1} (x-b) \leq 1 \},$$
for some **positive definite matrix  $M$**  and vector  $b$ .
- This helps us understand positive definite matrices. Consider  $f(x) = x^T M x$ , where  $M$  is positive definite. Its sub-level sets are  $\{ x \in \mathbb{R}^n : x^T M x \leq a \} = E(aM^{-1}, 0)$ .
- **Note:**  $E(M,b) = f(B)$  where  $f(x) = M^{1/2}x + b$ . So  $\text{vol } E(M,b) = |\det M^{1/2}| \cdot \text{vol } B = |\det M|^{1/2} \cdot \text{vol } B$ .



- Let  $M = \begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix}$
- Plot of level sets of  $x^T M x$ .

# Covering Hemispheres by Ellipsoids



- Let  $B = \{ \text{unit ball} \}$ .
- Let  $H = \{ x : x^T e_1 \geq 0 \} = \{ x : x_1 \geq 0 \}$ .
- Find a small ellipsoid  $B'$  that covers  $B \cap H$ .

# Rank-1 Updates

- **Def:** Let  $z$  be a column vector and  $\alpha$  a scalar. A matrix of the form  $I + \alpha z z^T$  is called a **rank-1 update matrix**.
- **Claim 1:** Suppose  $\alpha \neq -1/z^T z$ . Then  $(I + \alpha z z^T)^{-1} = I + \beta z z^T$  where  $\beta = -\alpha/(1 + \alpha z^T z)$ .
- **Claim 2:** If  $\alpha \geq -1/z^T z$  then  $I + \alpha z z^T$  is PSD.  
If  $\alpha > -1/z^T z$  then  $I + \alpha z z^T$  is PD.
- **Claim 3:**  $\det(I + \alpha z z^T) = 1 + \alpha z^T z$