

Today: The following are NP-hard:

1. Maximizing a convex function over a convex set.
2. Finding the point of largest norm in a polytope ($Ax \leq b$)
3. Computing a matrix p-norm, $\|M\|_p$, $2 < p < \infty$

All will follow from reductions from Max-Cut.

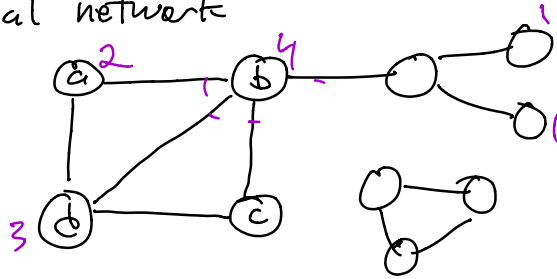
This is a graph problem.

Recall that a graph G , usually written $G = (V, E)$, consists of a set of vertices (aka "nodes"), V , and edges E connecting pairs of nodes.

Each edge is a subset of V of size 2, but we write the pair in parentheses, like (a, b)

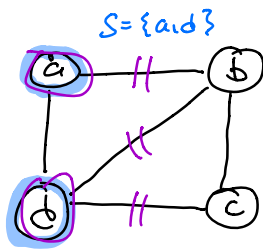
Ex 1 Social network

Ex 2

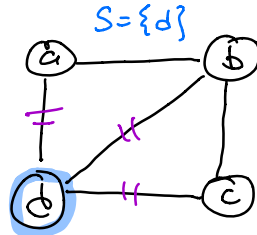


A "cut" is a subset of the vertices, $S \subseteq V$.

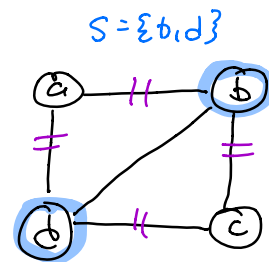
We count the size of the boundary of S , $\partial(S)$
= edges from inside to outside S



$$|\partial(S)| = 3$$



$$|\partial(S)| = 3$$



$$|\partial(S)| = 4$$

The maximum cut is $\text{maxcut}(G) = \max_{S \subseteq V} |\partial(S)|$

To make into a decision problem, we give G and k as input. yes if $\text{maxcut}(G) \geq k$

$(G, k) \in \text{MaxCut}$ if $\exists S \subseteq V$ st. $|\partial(S)| \geq k$

Answer is "yes" if G has a cut of size $\geq k$

MaxCut is in NP: S is the witness

And, it is NP-complete.

One can not even approximate its value.

→ Hastad '01 proved is NP-hard to even approximate within a factor $18/17$.

That is, for every problem $Y \in \text{NP}$,

\exists a polynomial time algorithm A that outputs G, k

i.e. $A(q) = (G, k)$

st. $\bullet q \in Y \Rightarrow G$ has a cut of size $\geq k$

$\bullet q \notin Y \Rightarrow$ all cuts in G are smaller than $\frac{17}{18} k$
 $\text{maxcut}(G) \leq \frac{17}{18} k$

The result is a little tighter than this: $\frac{17}{16} + \epsilon$, $\epsilon > 0$

Is a very deep and complicated result building on a decade of research by many luminaries.

Degree of a vertex is # of attached edges:

$$d_a = |\{b : (a,b) \in E\}|$$

Degree 3-Graphs: Berman & Karpinski '02

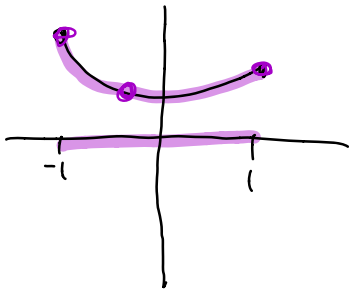
Reduce to graphs in which every vertex has degree 3

and

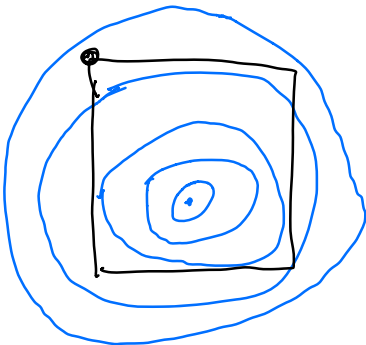
- $q \in \mathcal{Y} \Rightarrow \text{maxcut}(G) \geq k$
- $q \notin \mathcal{Y} \Rightarrow \text{maxcut}(G) \leq \frac{332}{333} k = \left(1 - \frac{1}{333}\right) k$

Will use this for p-norms.

Maximizing convex functions over convex sets

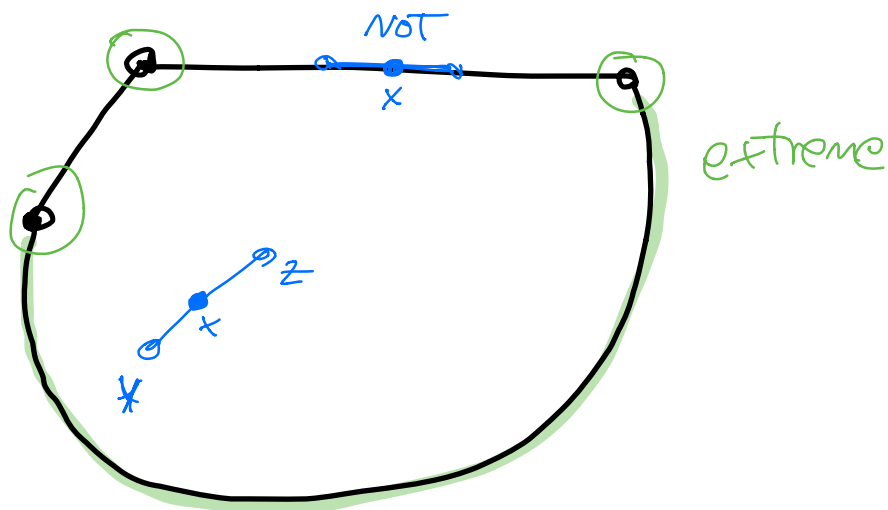


The maximum will be on the boundary, and values inside won't be very helpful.



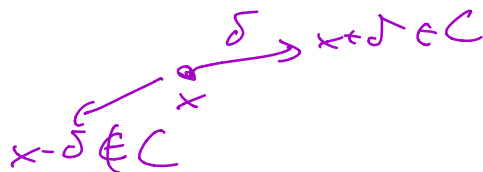
Claim: the maximum will be at a "corner".

Def x is an extreme point of a convex set C
 if there do NOT exist $y, z \in C$ and $\lambda \in (0, 1)$ s.t.
 $x = \lambda y + (1 - \lambda)z$.



Equivalently, for $\delta \neq 0$, $x + \delta \in C \Rightarrow x - \delta \notin C$.

When C is a polytope like $Ax \leq b$, the extreme points are the corners.



Idea of proof for $\{x: \|x\|_\infty \leq 1\}$ $-1 \leq x(i) \leq 1$

let x_* be opt. $y = (1, x_*(2), \dots, x_*(n))$

$z = (-1, x_*(2), \dots, x_*(n))$

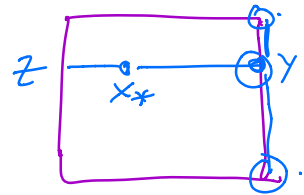
$x_* \in \gamma z$

Convexity $\Rightarrow f(x_*) \leq \max(f(y), f(z))$

so go to one of y or z

repeat for other coordinates

eventually make all coords ± 1 .



Proof. (for polytopes) let x_* be the maximum.

(Carathéodory \Rightarrow)

\exists corners v_0, \dots, v_d s.t. $x_* = \sum_{i=0}^d \lambda_i v_i$
 $\lambda_i \geq 0, \sum \lambda_i = 1$

Convexity $f(x_*) \leq \sum \lambda_i f(v_i)$

$\Rightarrow \exists v_i$ s.t. $f(x_*) \leq f(v_i)$

A convex relaxation of maxcut.

First, need to write algebraically.

Let $n = |V|$. For a set S , let $x_S(a) = \begin{cases} 1 & a \in S \\ -1 & a \notin S \end{cases}$

For $(a,b) \in E$

$$|x_S(a) - x_S(b)| = \begin{cases} 2 & \text{if } (a,b) \in \partial(S) \\ 0 & \text{otherwise, when } x_S(a) = x_S(b) \end{cases}$$

$$\frac{1}{4} \sum_{(a,b) \in E} (x_S(a) - x_S(b))^2 = |\partial(S)|$$

$$Q(x) = \frac{1}{4} \sum_{(a,b) \in E} (x(a) - x(b))^2$$

Is a sum of squares, so is convex.

Claim $\max_{\|x\|_\infty \leq 1} Q(x) = \text{maxcut}(G)$

So, have a polynomial time reduction from maxcut to problem of maximizing Q over $\|x\|_\infty \leq 1$, a convex set.

\Rightarrow Maximizing convex quadratics over $\|x\|_\infty \leq 1$ is NP hard.

The decision problem is: given Q, t , $\exists x: \|x\|_\infty \leq 1, Q(x) \geq t$?
Is in NP.

Proof $\max Q(x) \geq \text{maxcut}(G)$, $x = x_{S_*}$
 where $|\partial(S_*)| = \text{maxcut}(G)$
 $\max Q(x) = \text{maxcut}$. $\exists x_* \in \{\pm 1\}^n$ s.t. x_* is max
 $\exists S$ s.t. $x_* = x_S$
 $Q(x_*) = |\partial(S)| = \text{maxcut}(G)$

Note: Turned arbitrary maxcut problem
 into special convex quadratic.

If don't expect to solve maxcut, shouldn't expect
 to maximize convex quadratics.

It is a very special quadratic: the Laplacian

$$\underline{Q(x)} = \frac{1}{4} x^T L x \quad \text{where} \quad L(a,b) = \begin{cases} \text{degree of } a & \text{if } \underline{a=b} \\ -1 & \text{if } \underline{(a,b) \in E} \\ 0 & \text{o.w.} \end{cases}$$

$$\underline{(Lx)}(a) = d_a x(a) - \sum_{b: (a,b) \in E} x(b)$$

$$\begin{aligned} x^T L x &= \sum_a d_a x(a)^2 - \sum_a \sum_{b: (a,b) \in E} x(a)x(b) \\ &= \sum_{(a,b) \in E} (x(a)^2 + x(b)^2) - 2 \sum_{(a,b) \in E} x(a)x(b) \end{aligned}$$

$$\begin{aligned} &= \sum_{(a,b) \in E} (x(a) - x(b))^2 = 4Q(x) \\ & \quad Q(x) = \frac{1}{4} \sum (x(a) - x(b))^2 \end{aligned}$$

(9.10) EE

Thm The problem $\max_x \|x\|_2^2$ s.t. $Ax \leq b$ is NP-hard

proof We will reduce the problem

(1) $\max_x x^T L x$ s.t. $\|x\|_\infty \leq 1$ to this one

Consider $x^T (L+I) x = x^T L x + \underbrace{x^T x}_n = n$

(2) $= \max_x x^T (L+I) x$ s.t. $\|x\|_\infty \leq 1 = (1) + n$

$L+I$ is pos def $\Rightarrow \exists$ invertible, symmetric B s.t.
 $B^T B = L+I$

$$x^T (L+I) x = x^T B^T B x$$

$$y = Bx \text{ and } x = B^{-1}y$$

(2) $\Leftrightarrow \max_y \underbrace{y^T y}_{=n}$ s.t. $\|B^{-1}y\|_\infty \leq 1$ |

$\|B^{-1}y\|_\infty \leq 1$ let $b_1 \dots b_n$ be rows of B^{-1}

$$\text{get } b_i^T y \leq 1 \quad -b_i^T y \leq 1$$

In fact, hard to approximate.

Matrix Norms for $p > 2$.

$$\|M\|_p = \max_x \frac{\|Mx\|_p}{\|x\|_p} = \max_{\|x\|_p \leq 1} \|Mx\|_p$$
$$\|x\|_p = \left(\sum_i x(i)^p \right)^{1/p}$$

$p=2$ is max singular value

$p=\infty$ is $\max_i \|M(i, \cdot)\|_1$

$p=1$ is $\max_i \|M(\cdot, i)\|_1$

after we correct
minor mistakes

For all other p is NP-hard.

Bhaskara & Vijayaraghavan show is NP-hard
to approximate within any constant.

Write $(M, t) \in$ Matrix p -norm if $\|M\|_p \geq t$

Thm For all $p > 2$, $c > 1$, and $Y \in NP$,

\exists poly time algorithm A that outputs M, t s.t.
for $A(\varphi) = (M, t)$

$$\varphi \in Y \Rightarrow \|M\|_p \geq \underline{ct}$$

$$\varphi \notin Y \Rightarrow \|M\|_p \leq \underline{t}$$

A depends upon c .

How prove this for all $c > 1$?

First show for some $c > 1$, and then amplify.

$$A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix} \quad x \otimes y = \begin{pmatrix} x_1 \cdot y_1 \\ x_1 \cdot y_2 \\ \vdots \\ x_1 \cdot y_n \\ x_2 \cdot y_1 \\ \vdots \\ x_n \cdot y_n \end{pmatrix} \in \mathbb{R}^{n^2}$$

$x, y \in \mathbb{R}^n$

Recall the Kronecker product:

$$A \otimes B = \begin{pmatrix} a_{11} B & a_{12} B & \dots & a_{1n} B \\ \vdots & & & \\ a_{n1} B & \dots & & a_{nn} B \end{pmatrix}$$

$$\begin{pmatrix} x_1 y_1 \\ x_2 y_1 \\ \vdots \\ x_n y_1 \\ x_1 y_2 \\ \vdots \\ x_n y_2 \\ \vdots \\ x_1 y_n \\ \vdots \\ x_n y_n \end{pmatrix} = \text{vec}(x y^T)$$

$x_i y_j$

Thm $\|A \otimes B\|_p = \|A\|_p \cdot \|B\|_p$

Kron

proof sketch:

1. For vectors x, y $\|x \otimes y\|_p = \|x\|_p \cdot \|y\|_p$

$$\boxed{(A \otimes B)(x \otimes y) = (Ax) \otimes (By)}$$

$$\Rightarrow \|A \otimes B\|_p \geq \|A\|_p \cdot \|B\|_p$$

let $x_* \in \arg \max_x \|Ax\|_p / \|x\|_p$, y_* for B

2. $A \otimes B = (A \otimes I)(I \otimes B)$

$$\|A \otimes B\|_p = \|A \otimes I\|_p \cdot \|I \otimes B\|_p = \|A\|_p \|B\|_p$$

$$I \otimes B = \begin{pmatrix} B & & 0 \\ & B & \\ 0 & & \ddots \\ & & & B \end{pmatrix} \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} B y_1 \\ \vdots \\ B y_n \end{pmatrix}$$

So, it is hard to distinguish $\|M\|_p \leq s$ from $\geq b$
 it is hard to distinguish $\|M \otimes M\|_p \leq s^2$ from $\geq b^2$

This squares the inapproximability factor.

Can do this any constant number of times,
 and still be a polynomial time reduction.

To get some constant:

Idea: want $\sum_{(a,b) \in E} |x(a) - x(b)|^p$ $\|x\|_p = 1$

But, can't force $x(a) = \pm 1$.

So, add terms to penalize $|x(a)| \notin (1-\epsilon, 1+\epsilon)$

Claim 1: let $F(x) = \frac{|x+1|^p + |x-1|^p}{1 + |x|^p}$ $x = -1 \quad \frac{0 + 2^p}{1+1}$

a. $F(\pm 1) = 2^{p-1}$

b. $\forall x, F(x) \leq 2^{p-1}$

c. $\forall \epsilon > 0 \exists \delta > 0$ s.t. $|x| \notin (1-\epsilon, 1+\epsilon)$
 $\Rightarrow F(x) \leq 2^{p-1} - \delta$

Claim 2: $\forall \varepsilon > 0 \exists C > 0$ s.t.

$$\text{for } F(x, y) = \frac{|x-y|^p + C[|x+1|^p + |x-1|^p + |y+1|^p + |y-1|^p]}{2 + |x|^p + |y|^p}$$

if $xy < 0$ and $|x|, |y| \in (1-\varepsilon, 1+\varepsilon)$

$$F(x, y) \leq C 2^{p-1} + \frac{(1+\varepsilon)^p 2^{p-1}}{2 + |x|^p + |y|^p}$$

if $xy \geq 0$, $|x| \notin (1-\varepsilon, 1+\varepsilon)$ or $|y| \notin (1-\varepsilon, 1+\varepsilon)$

$$F(x, y) \leq C 2^{p-1} + \frac{(2\varepsilon)^p 2^{p-1}}{2 + |x|^p + |y|^p} \quad \text{smaller}$$

For a 3-regular graph G , set

$$g(x) = \frac{\sum_{(a,b) \in E} |x(a) - x(b)|^p + 3C[|1+x(a)|^p + |1-x(a)|^p + |1+y(a)|^p + |1-y(a)|^p]}{3n + 3 \sum_a |x(a)|^p}$$

lem let $\text{maxcut}(G) = |\partial(S)| = \gamma \frac{3}{2} n$.

$$g(X_S) = C 2^{p-1} + \gamma 2^{p-2}, \text{ and}$$

$$\forall x \quad g(x) \leq C 2^{p-1} + (2\varepsilon)^p 2^{p-2} + \gamma (1+\varepsilon)^p (1-\varepsilon)^p 2^{p-2}$$

For any $0 < \gamma_0 < \gamma_1$, is a small enough ε so that computing $\max_x g(x)$ allows one to distinguish

$$\text{maxcut}(G) \leq \gamma_0 \frac{3}{2} n \quad \text{from} \quad \geq \gamma_1 \frac{3}{2} n$$

So, is NP-hard to compute $\max_x g(x)$ to within some constant

To turn into a matrix, introduce new variable $x(0)$, which want to have value $n^{1/p}$ so $x(0)/n^{1/p} = 1$, and consider

$$\frac{\sum_{(a,b) \in E} |x(a) - x(b)|^p + 3C \left[\left| \frac{x(0)}{n^{1/p}} - x(a) \right|^p + \left| \frac{x(0)}{n^{1/p}} + x(a) \right|^p + \left| \frac{x(0)}{n^{1/p}} - x(b) \right|^p + \left| \frac{x(0)}{n^{1/p}} + x(b) \right|^p \right]}{3|x(0)|^p + 3 \sum_a |x(a)|^p}$$

Show its max value is same as max of g .

Denominator = $3 \|x\|_p^p$ and

Numerator = $\|Mx\|_p^p$ for some matrix M .

So, is hard to compute $\max_x \frac{\|Mx\|_p}{\|x\|_p} = \|M\|_p$,

and to approximate it up to some constant.