

Universal Rigidity of Generic Frameworks

Steven Gortler

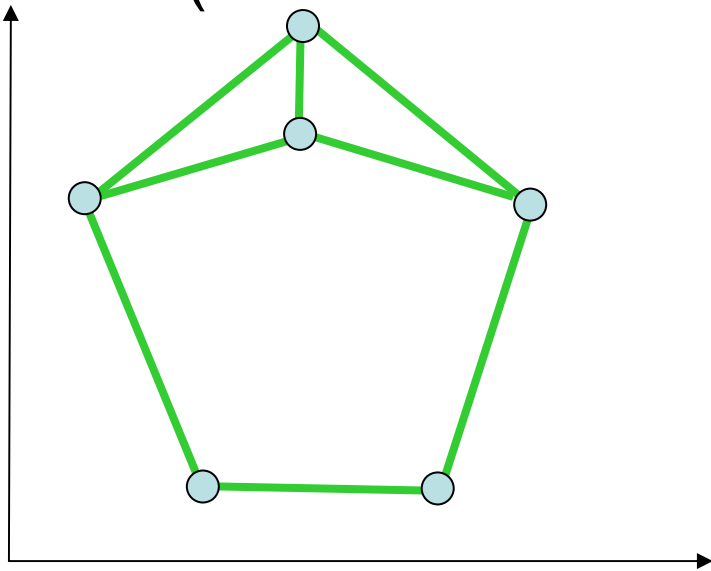
joint with Dylan Thurston

Degrees of Graph Rigidity



Framework of G

- Given a graph: “ G ”,
 - v vertices, e edges
- Framework p : drawing in E^d
 - (Don't care about edge-edge crossing)

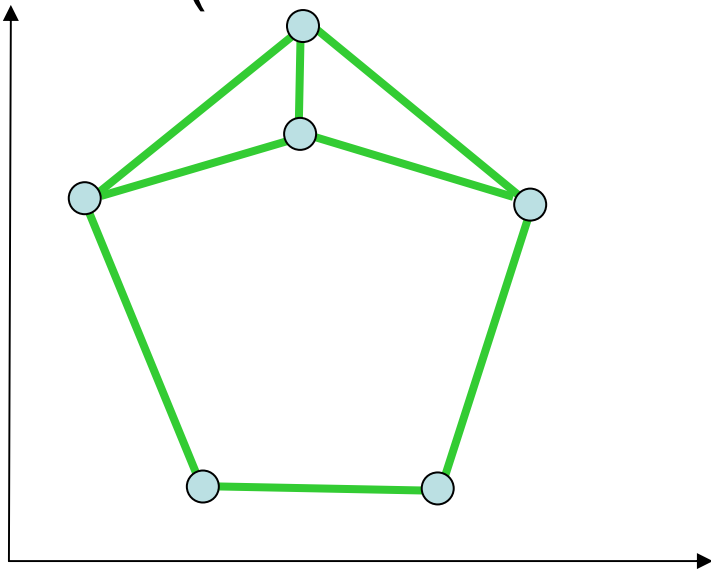


Generic

- We will restrict ourselves to frameworks that are generic in E^d
- Think “randomly perturbed” in E^d
- No algebraic coincidences
- Coincidences ruin the thms.
 - Quite typical scenario for rigidity

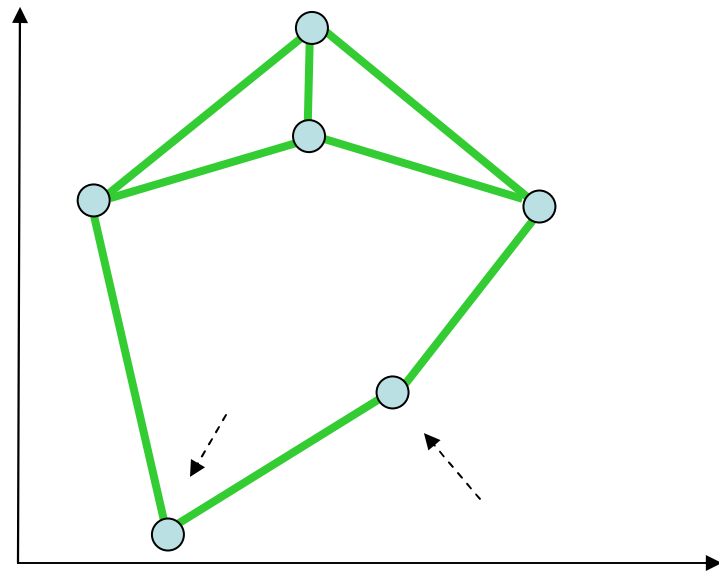
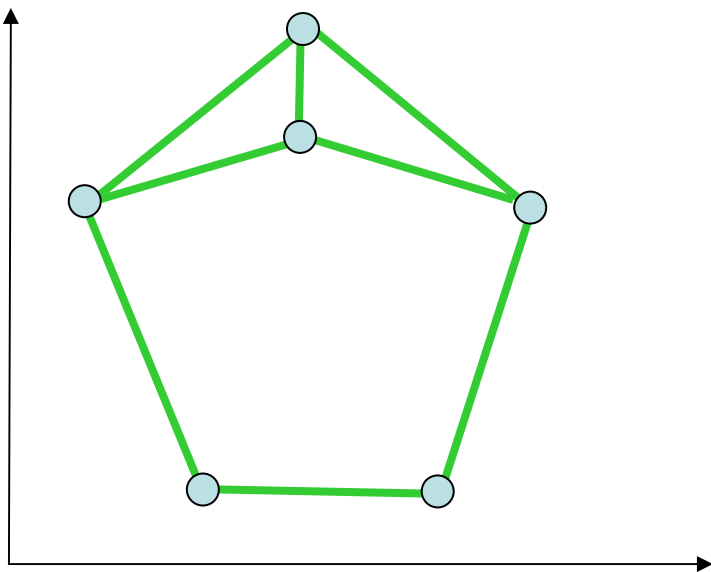
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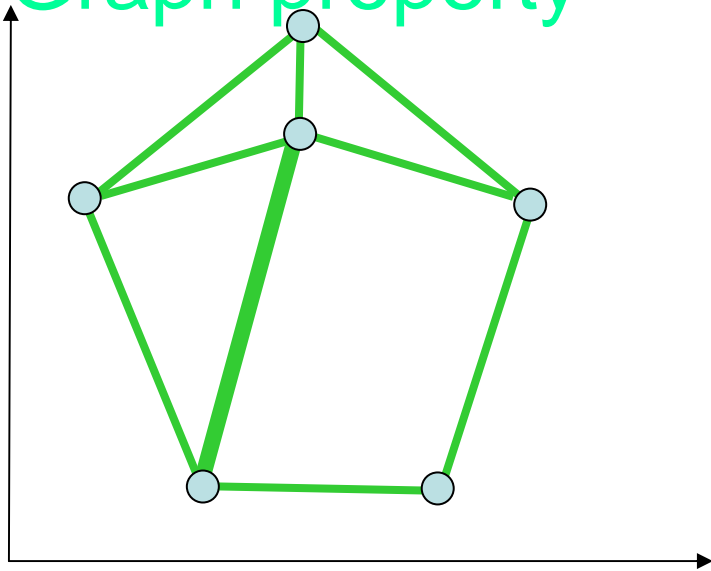
Locally flexible

- Can find continuum of frameworks in d -dims with same edge lengths
- Graph property (for generic frameworks)



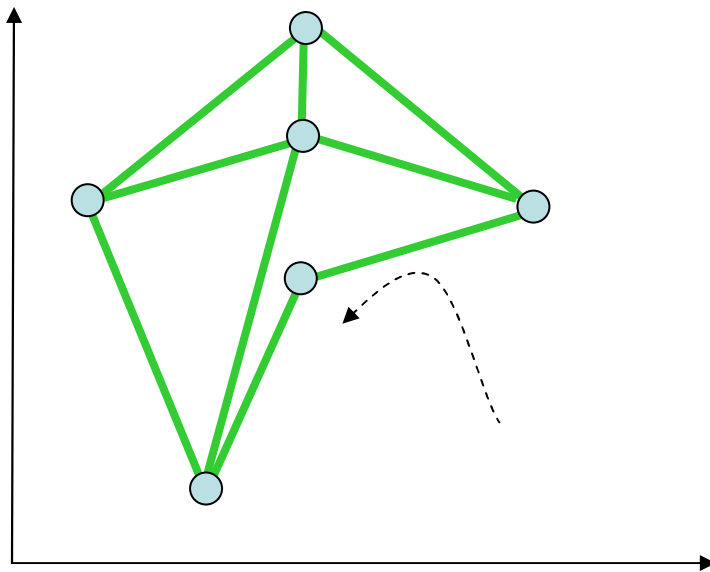
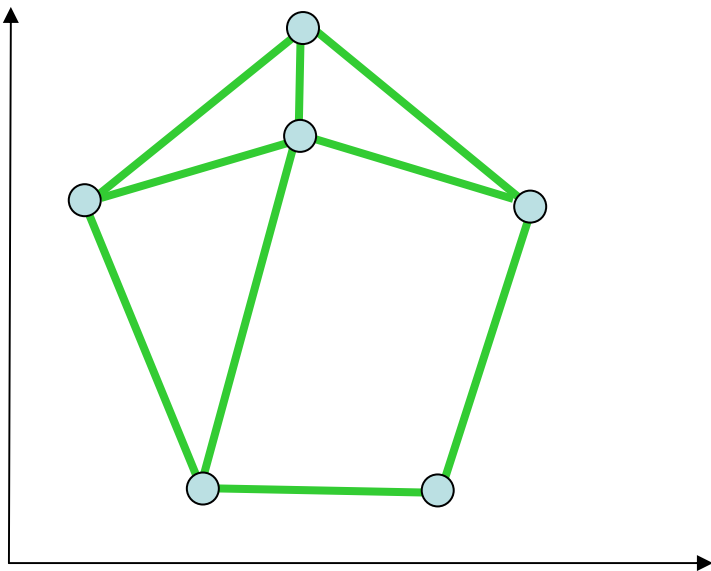
Locally rigid

- No continuum
- Well understood
 - Rank of appropriate matrix
- Graph property



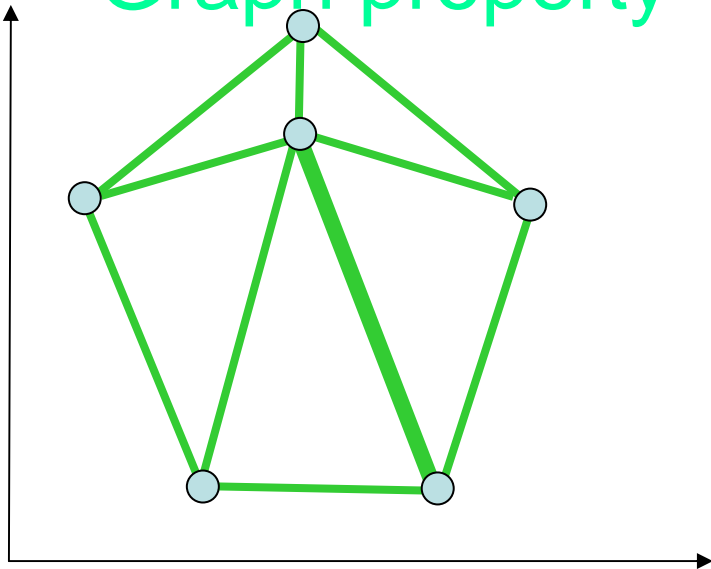
Locally rigid

- No continuum
- But can find other discrete framework in d-dims



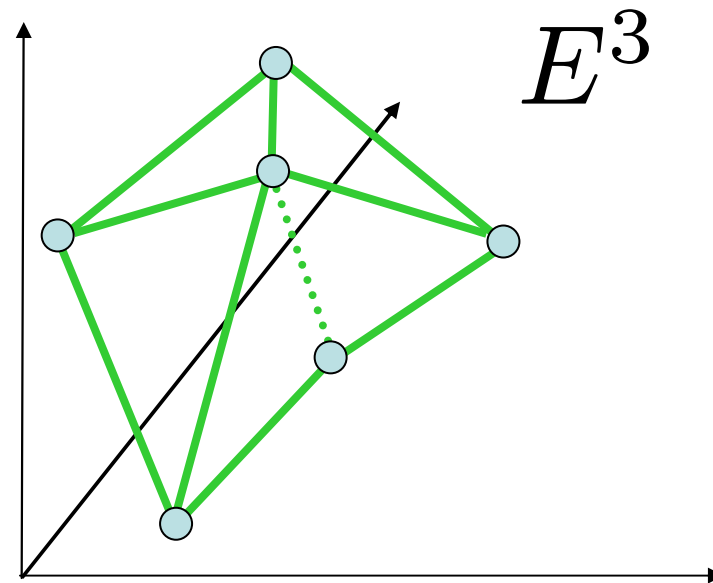
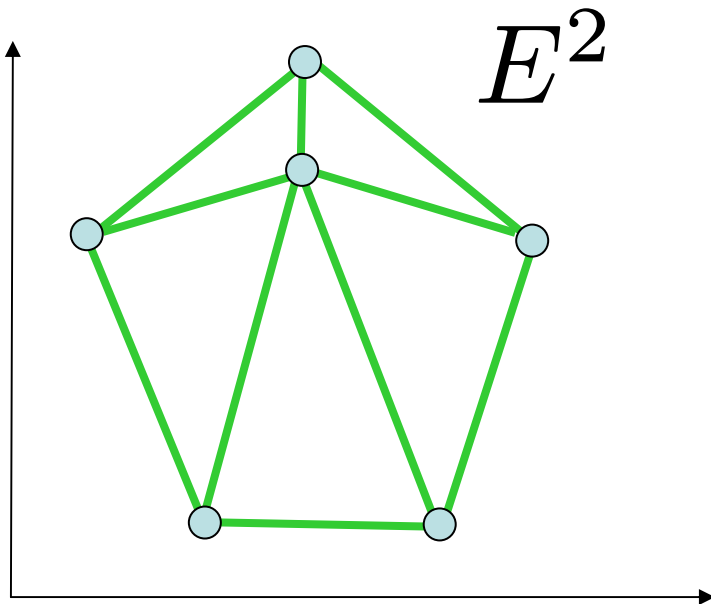
Globally Rigid

- No other framework in d-dims
- Well understood [C'82, GHT '07]
 - Rank of appropriate matrix
- Graph property



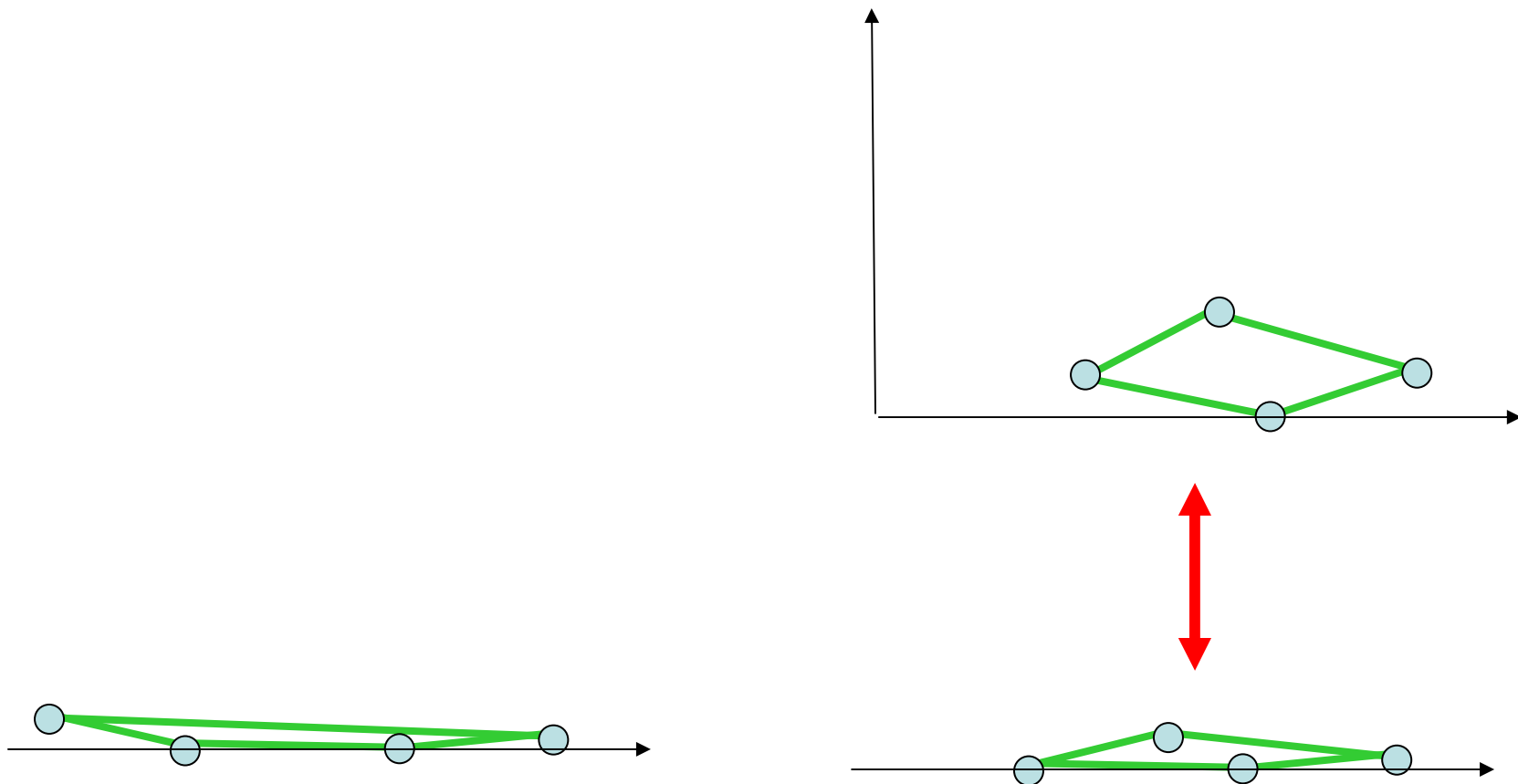
Globally Rigid

- No other framework in d-dims
- But can be other frameworks in higher dims

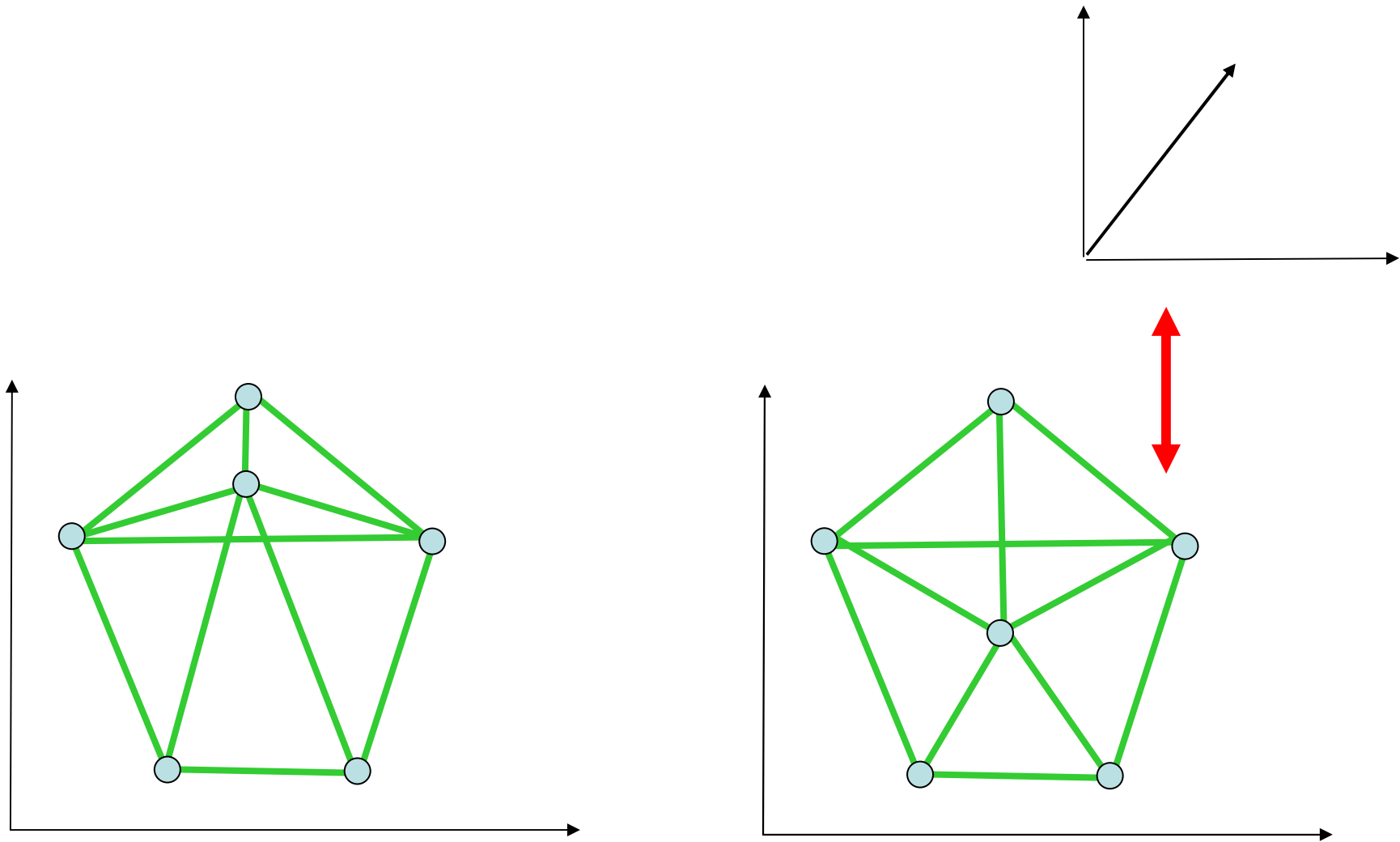


UR: can depend on the generic embedding: 1D

- Unlike local and global rigidity

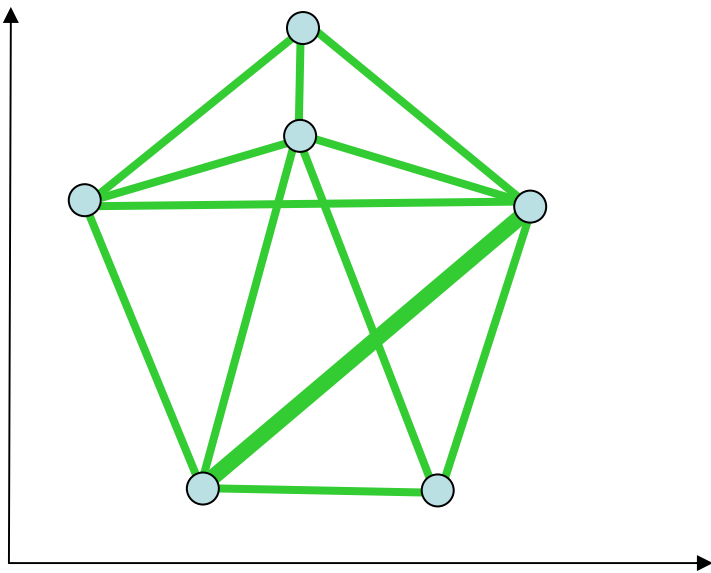


UR: can depend on the generic
embedding: 2D

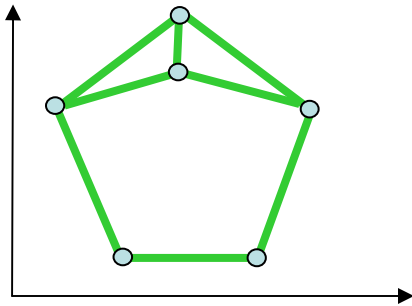


Generically UR Graphs

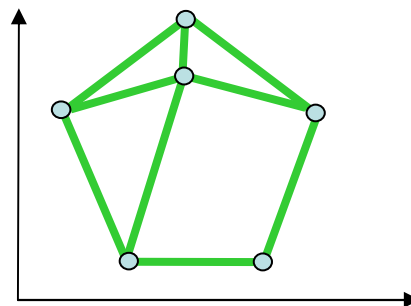
- Graphs such that ALL d -dim generic frameworks are UR



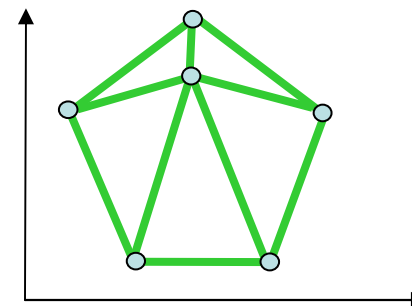
Degrees of rigidity



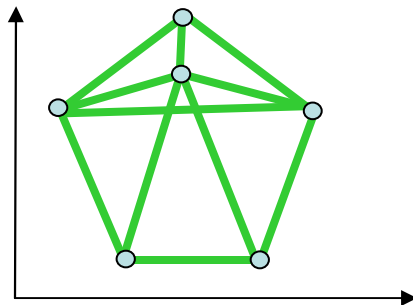
Locally Flexible graph



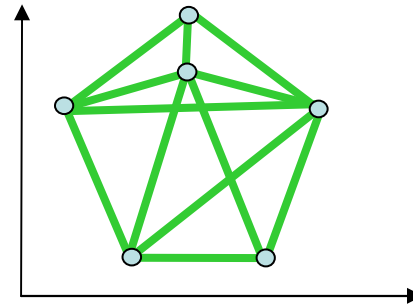
Locally Rigid graph



Globally Rigid graph

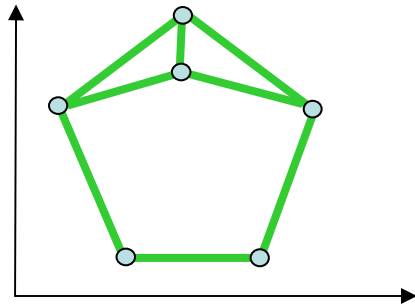


Universally Rigid Framework

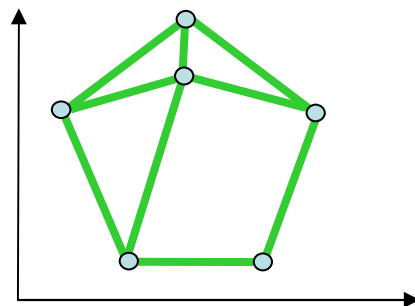


Universally Rigid Graph

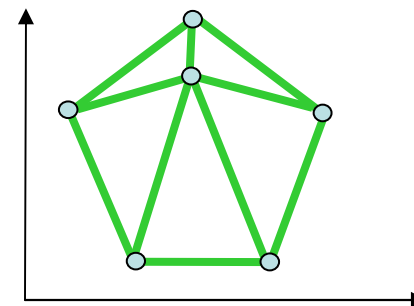
Degrees of rigidity



Locally Flexible graph

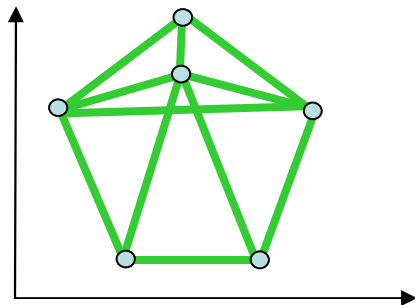


Locally Rigid graph

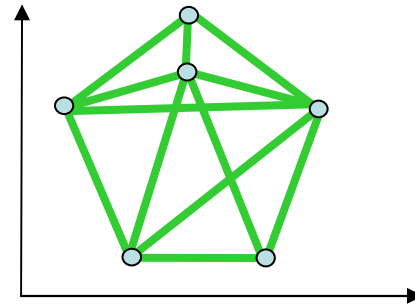


2007

Globally Rigid graph

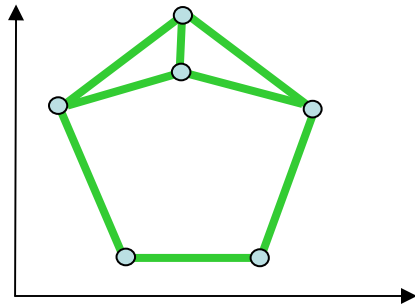


Universally Rigid Framework

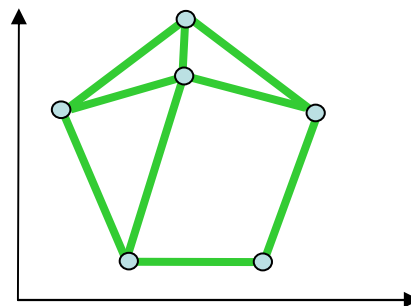


Universally Rigid Graph

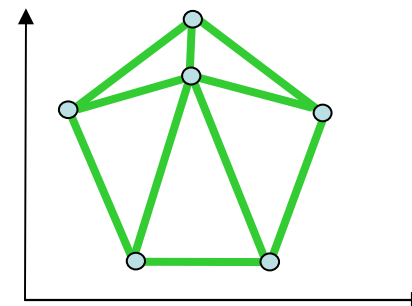
Degrees of rigidity



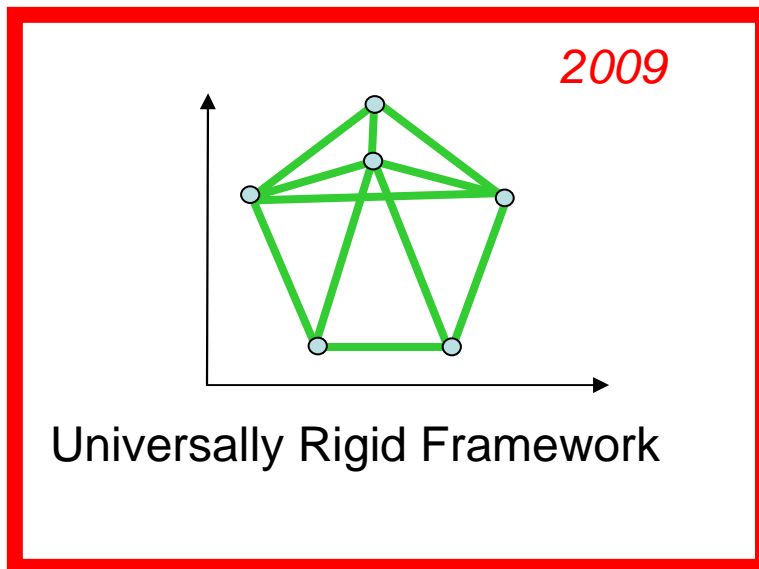
Locally Flexible graph



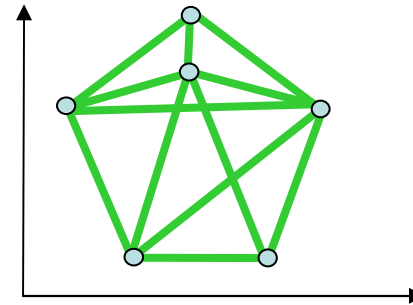
Locally Rigid graph



Globally Rigid graph



Universally Rigid Framework



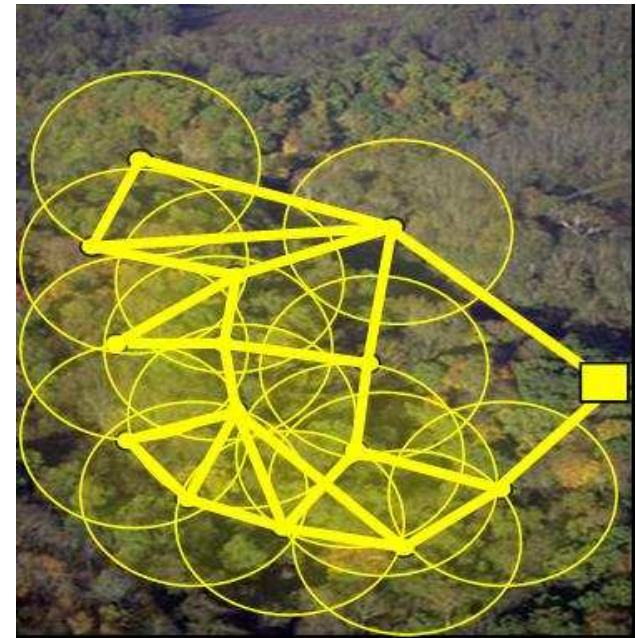
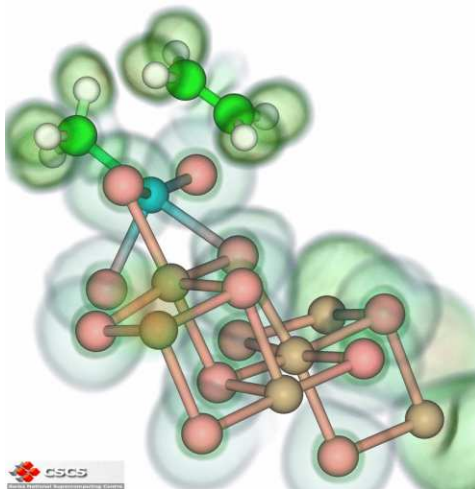
Universally Rigid Graph

Motivation for UR



Motivation for UR

- General embeddings from distances is NP-hard
- But it is useful in application
 - Molecular structure from NMR data
 - Sensor networks



Motivation for UR

- Try semidefinite programming (SDP) [LLR '95]
 - Unknown Gram matrix
 - Positive semidefinite (PSD) constraint
 - Length squared constraints are linear
- Will get solution in very high dimension
- Rank constraints are non-convex
 - UR: No need for explicit rank constraint
 - [So Ye '07]
- Wish to characterize this class

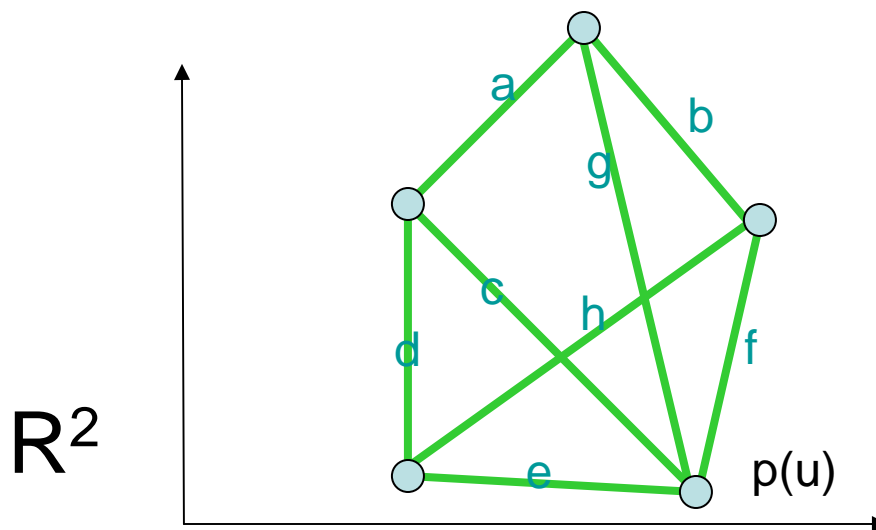
Define terms



Equilibrium Stress Vector (ESV) of p

- A real number w_{uv} on each edge e_{uv}

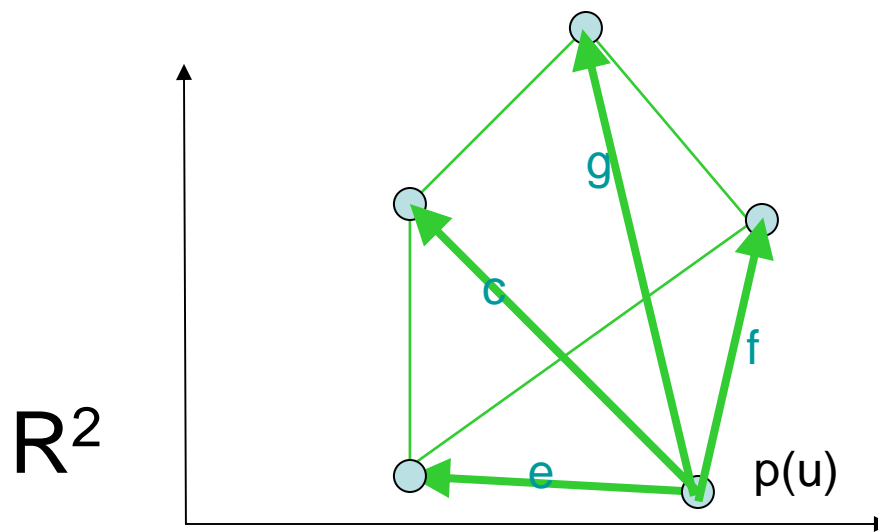
$$\forall u \sum_v \omega_{uv} [\rho(v) - \rho(u)] = \vec{0}$$



ESV of p

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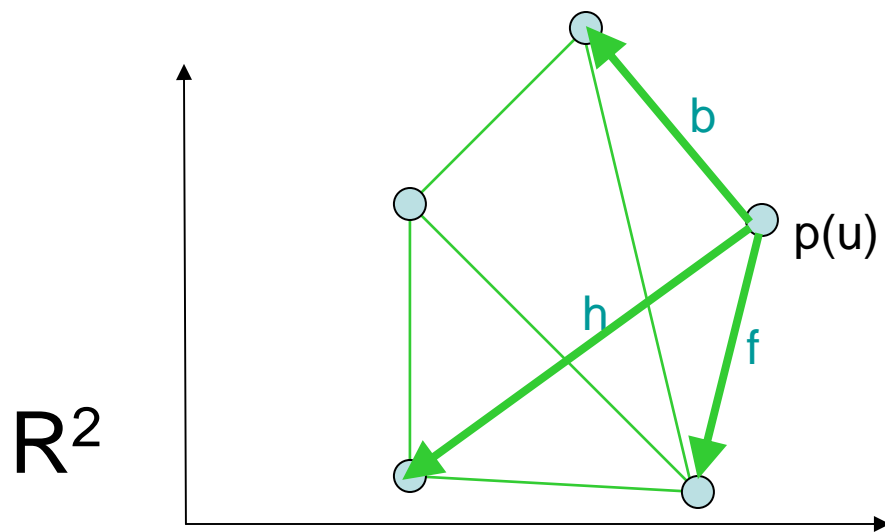
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

























































Equilibrium Stress Matrix (ESM) of p

- v by v matrix
- $\Omega_{uv} := \Omega_{vu} := \omega_{uv}$
- $\Omega_{uv} := 0$ for non-edges
- $\Omega_{uu} := -\sum_v \omega_{uv}$ for diagonals
- “Matricize the ESV”

Rank

OFFICERS INSIGNIA OF THE UNITED STATES ARMED FORCES

O-1	O-2	O-3	O-4	O-5	O-6	O-7	O-8	O-9	O-10	SPECIAL
NAVY										
   ENSIGN	   LIEUTENANT JUNIOR GRADE	   LIEUTENANT	   LIEUTENANT COMMANDER	   COMMANDER	   CAPTAIN	   COMMODORE ADMIRAL*	   REAR ADMIRAL (O-7 & O-8)	   VICE ADMIRAL	   ADMIRAL	   FLEET ADMIRAL
MARINES										
 SECOND LIEUTENANT	 FIRST LIEUTENANT	 CAPTAIN	 MAJOR	 LIEUTENANT COLONEL	 COLONEL	 BRIGADIER GENERAL	 MAJOR GENERAL	 LIEUTENANT GENERAL	 GENERAL	
ARMY										
 SECOND LIEUTENANT	 FIRST LIEUTENANT	 CAPTAIN	 MAJOR	 LIEUTENANT COLONEL	 COLONEL	 BRIGADIER GENERAL	 MAJOR GENERAL	 LIEUTENANT GENERAL	 GENERAL	 GENERAL OF THE ARMY
AIR FORCE										
 SECOND LIEUTENANT	 FIRST LIEUTENANT	 CAPTAIN	 MAJOR	 LIEUTENANT COLONEL	 COLONEL	 BRIGADIER GENERAL	 MAJOR GENERAL	 LIEUTENANT GENERAL	 GENERAL	 GENERAL OF THE AIR FORCE

Equilibrium Stress Matrix (ESM) of p

- By construction, kernel must contain:
 - Each of the d -coordinates of p

$$\Omega \rho^j$$

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$$\Omega \rho^j = \sum_v \omega_{uv} [\rho^j(v) - \rho^j(u)] = 0$$

$$\Omega \vec{1} = \sum_v \omega_{uv} [1 - 1] = 0$$

Equilibrium Stress Matrix (ESM) of p

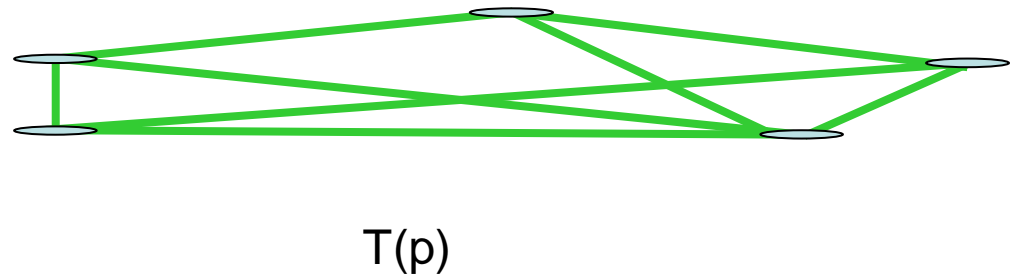
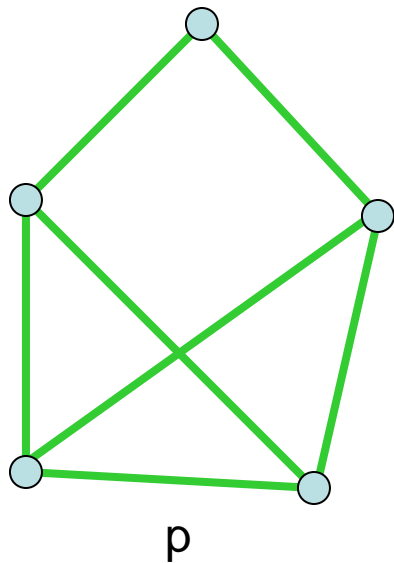
- By construction, kernel must contain:
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- Maximal possible rank is $v-d-1$

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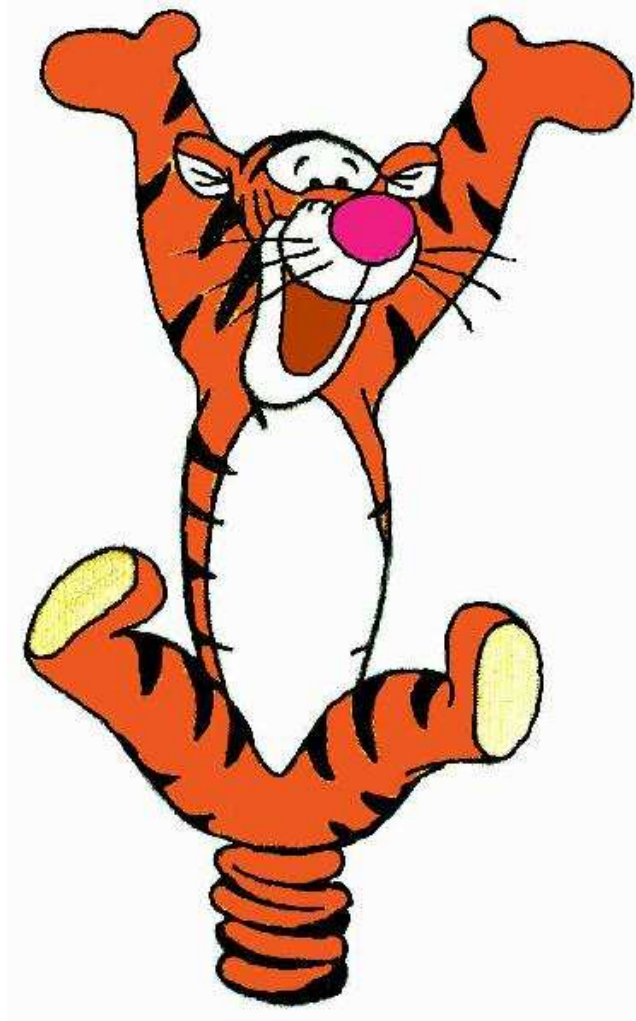
$$\Omega \vec{1} = \sum_v \omega_{uv} [1 - 1] = 0$$

Maximal possible rank is $v-d-1$

- Any affine transform ' $T(p)$ ' will satisfy **any** stress w of p
- If Ω has maximal rank, then these are the **only** satisfying drawings



Springs and Struts

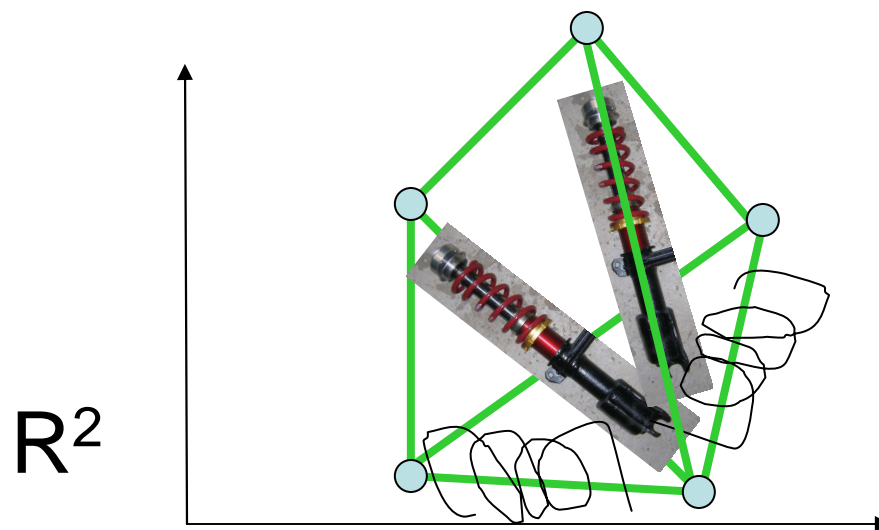


Stress Energy

- Stress vector -> a quadratic spring/strut energy

$$E(\rho) := \sum_{uv} \omega_{uv} |\rho(v) - \rho(u)|^2$$

- ESV condition == p is an equilibrium point of this energy
 - Max/Min/Saddle

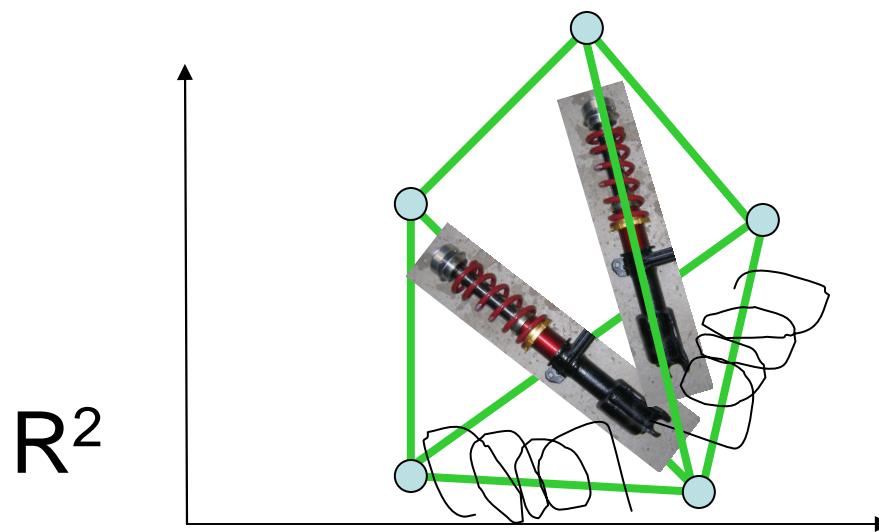


Stress Energy

- Stress vector -> a quadratic spring/strut energy

$$E(\rho) := \sum_{uv} \omega_{uv} |\rho(v) - \rho(u)|^2 = \sum_j (\rho^j)^t \Omega \rho^j$$

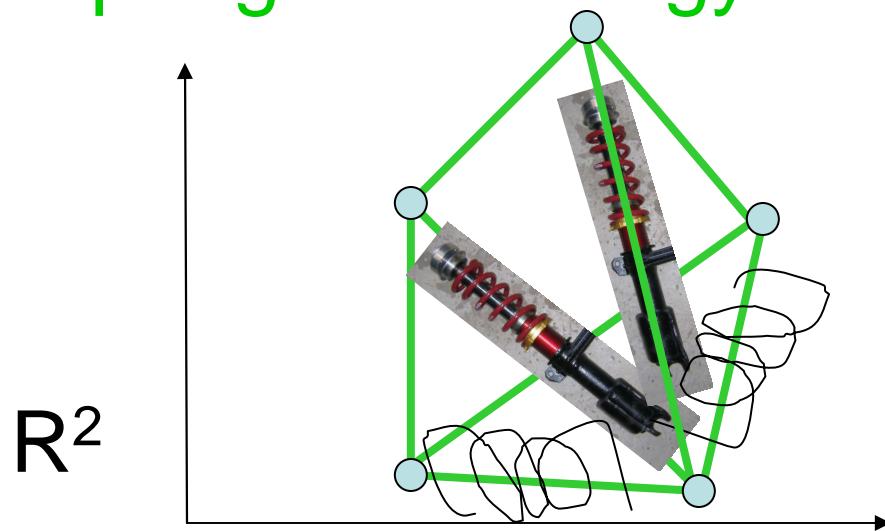
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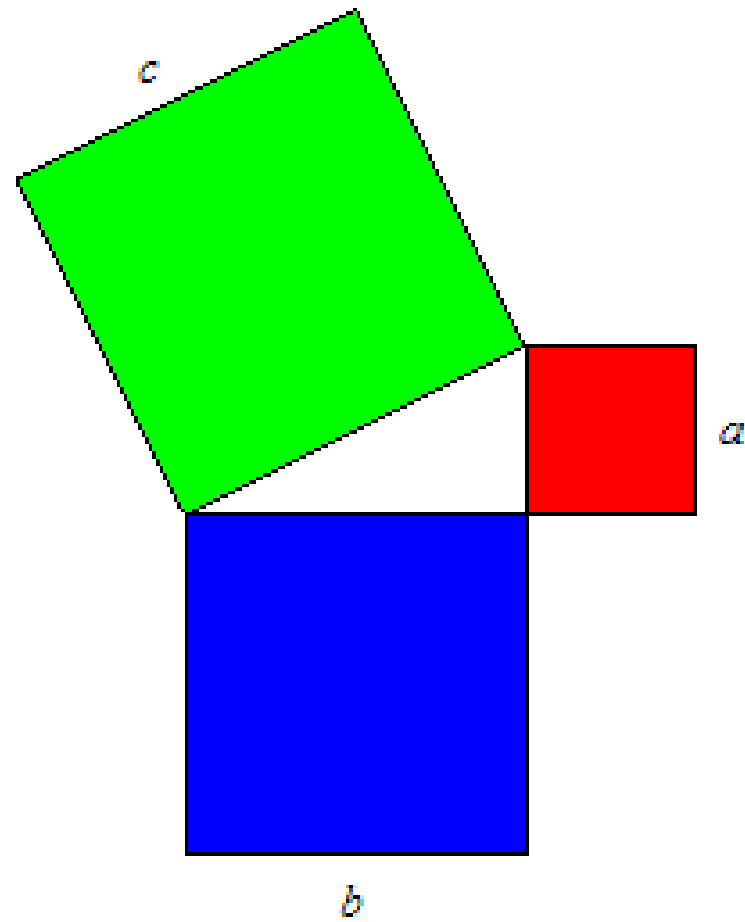
Positive semidefinite

$$E(\rho) := \sum_j (\rho^j)^t \Omega \rho^j$$

- If Ω is positive semidefinite, then p is in fact a min of the spring/strut energy



Theorems



Sufficiency [Con 82]

If p is generic in E^d

p has a ESM: Ω of rank $v-d-1$.

Ω is PSD

Then

p is UR

Main result today [GT 09]

If p is generic in E^d

p is UR

Then

p has a ESM: Ω of rank $v-d-1$.

Ω is PSD

Characterization

If p is generic in E^d

p is UR

==

p has a ESM: Ω of rank $v-d-1$.

Ω is PSD

Compare to GGR [Con 89+ GHT 07]

If p is generic in E^d
 ~~p is UR GR in E^d~~

==

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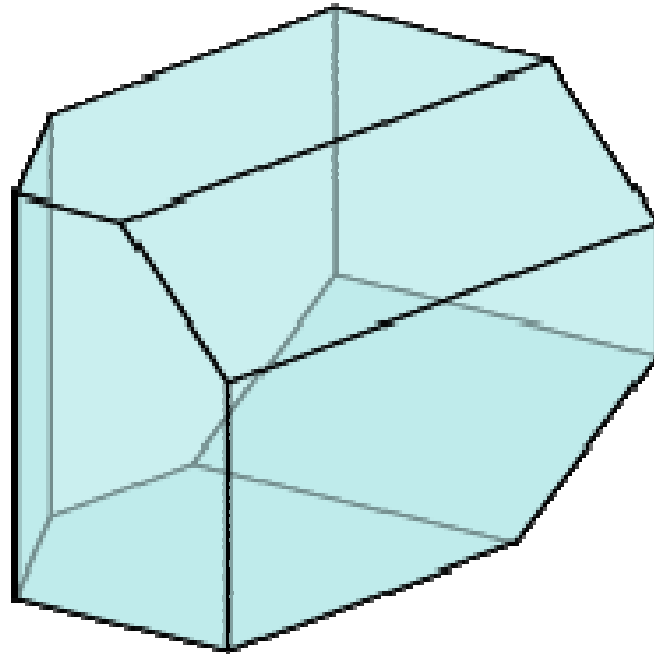
~~Ω is PSD~~

Proof



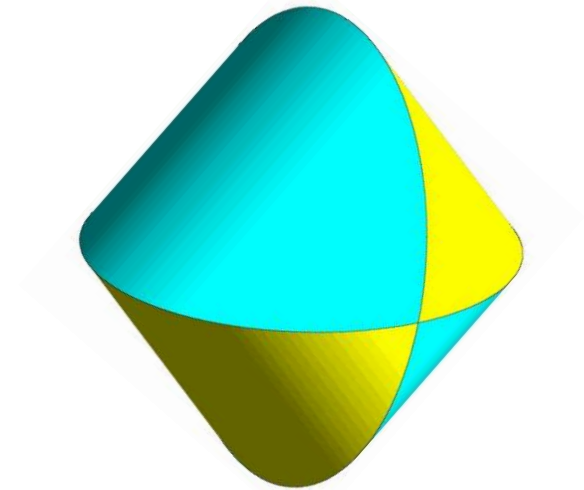
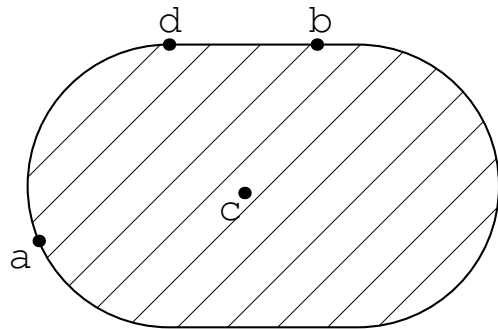
Convex preliminaries

- Stratify by “dimension”
- Obvious for polytopes
- Two notions in general



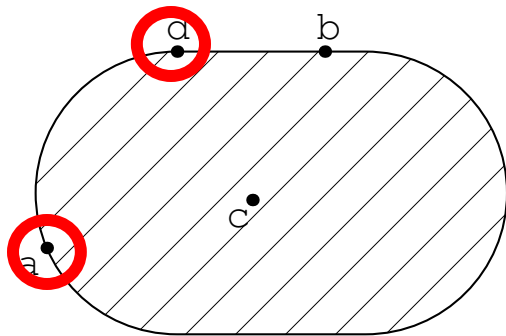
Convex extremity

k -extreme: not interior to $k+1$ simplex in the body

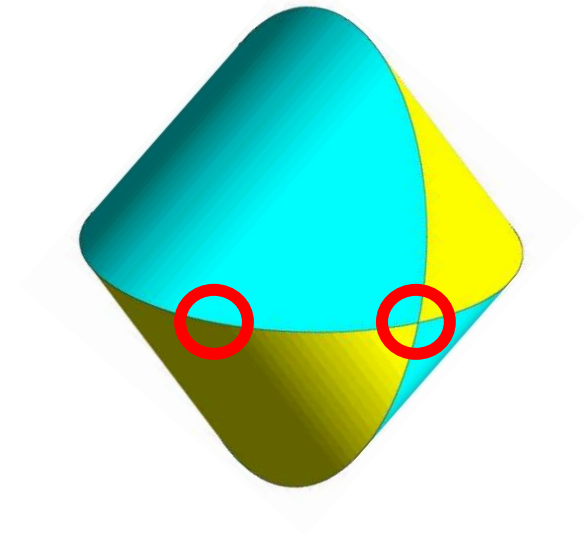


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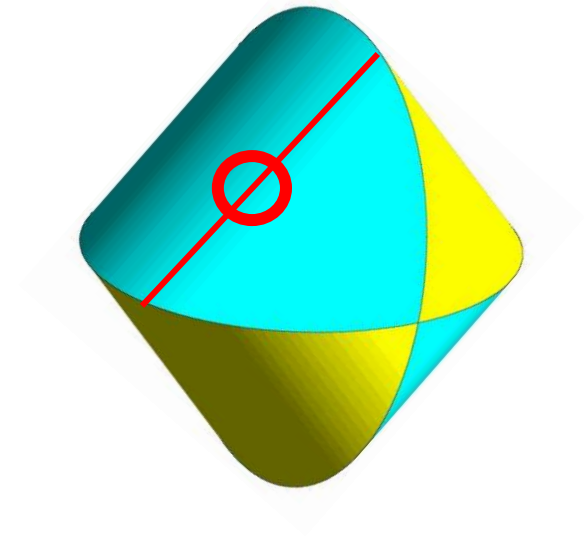
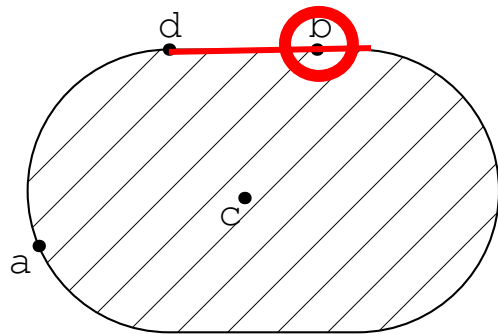


0-extreme



Convex extremity

k -extreme: not interior to $k+1$ simplex in the body



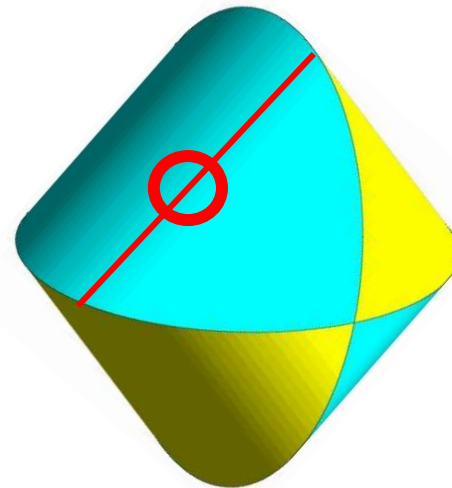
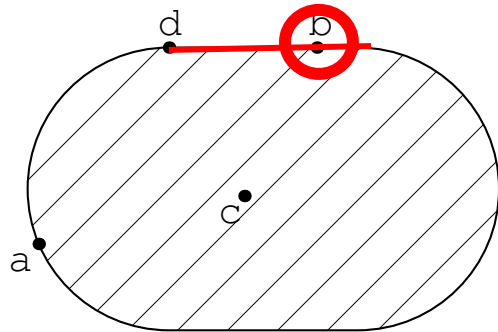
1-extreme

Convex extremity

k -extreme: not interior to $k+1$ simplex in the body

$\text{Face}(x)$: largest convex subset with x in interior

($\text{Dim Face}(x) = \text{smallest } k \text{ st } x \text{ is } k\text{-extreme}$)



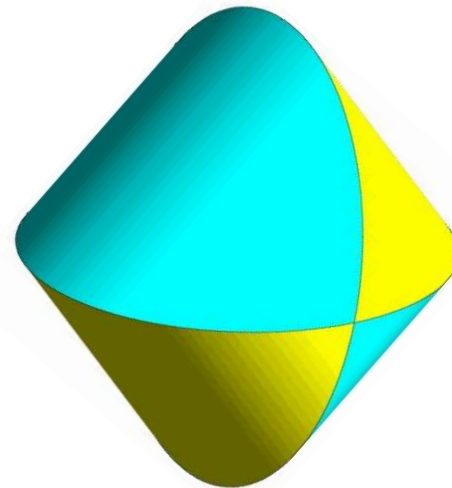
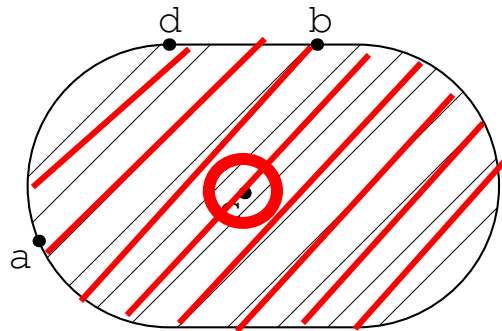
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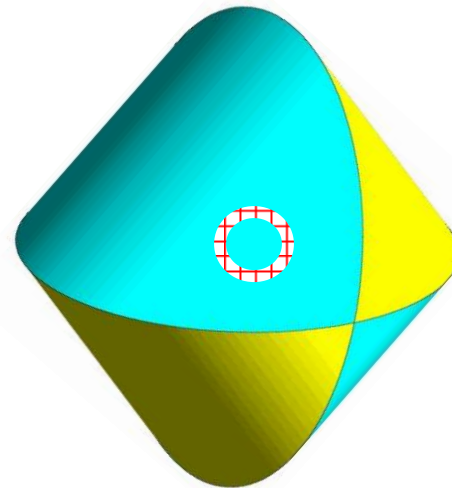
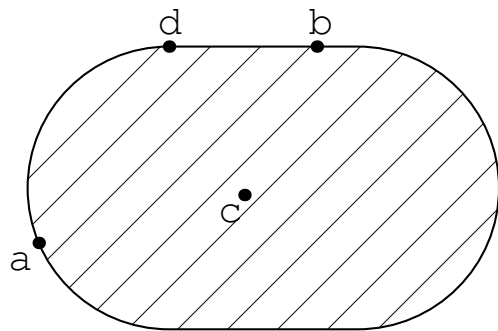
2-extreme

Convex extremity

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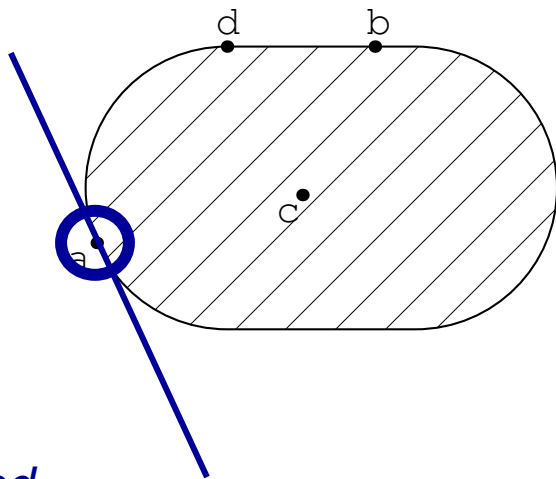
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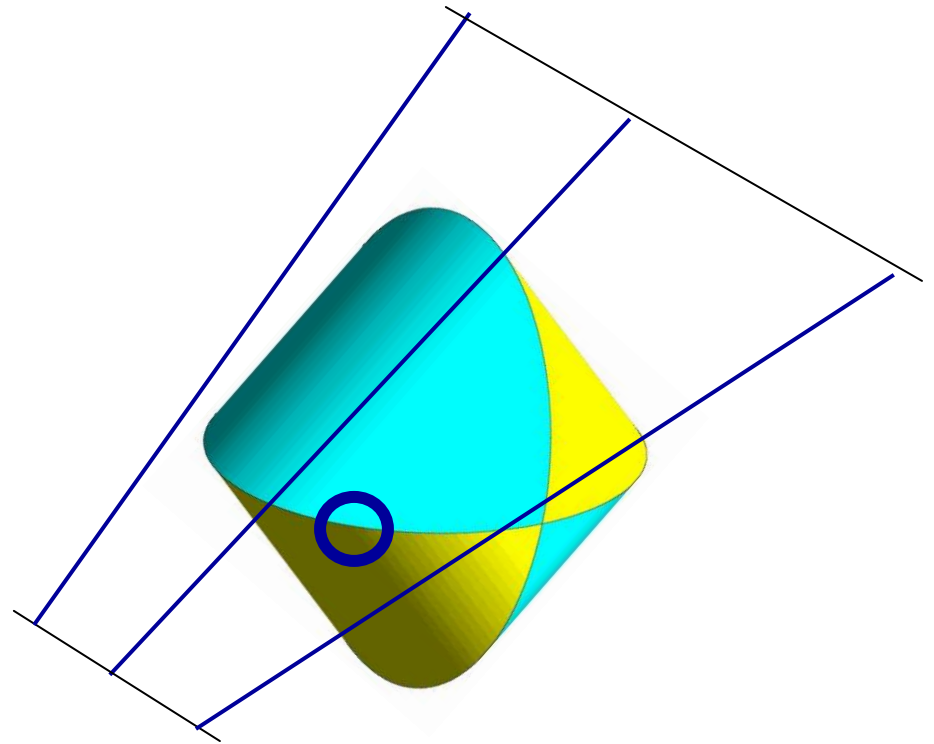
3-extreme

Convex exposedness

- k -exposed: exists halfspace with k dim intersection

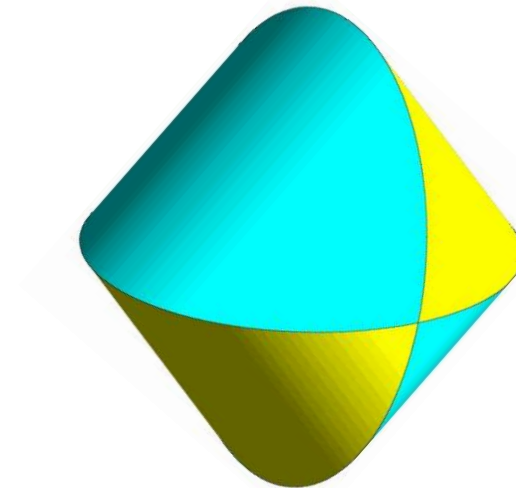
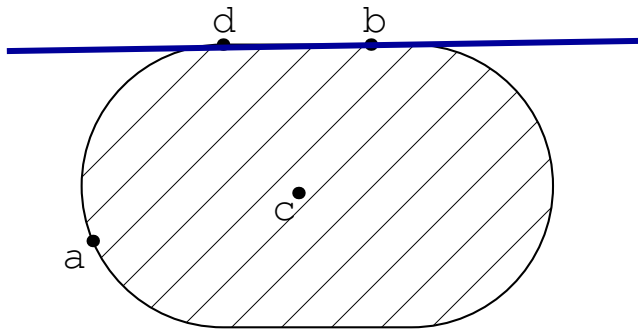


0-exposed



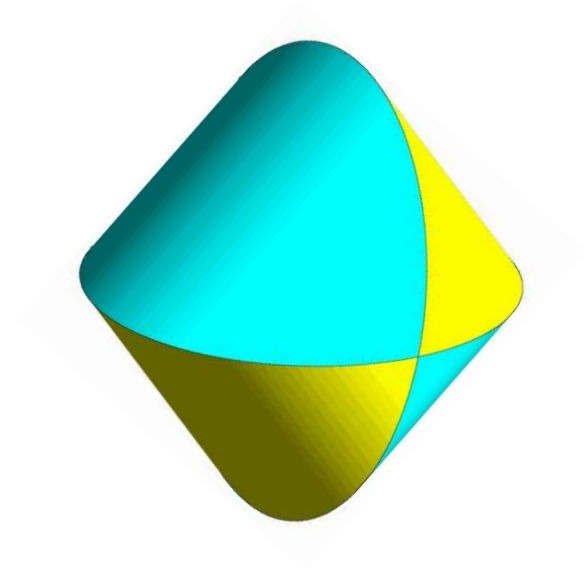
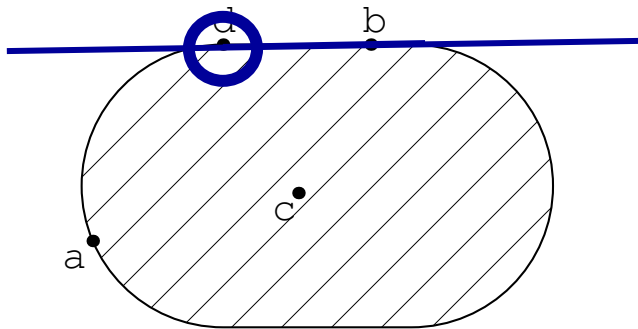
Convex exposedness

- Point d is 0-extreme (with 0-dim face)
- But not 0-exposed!
 - Every supporting hyperplane includes more stuff



Convex exposedness

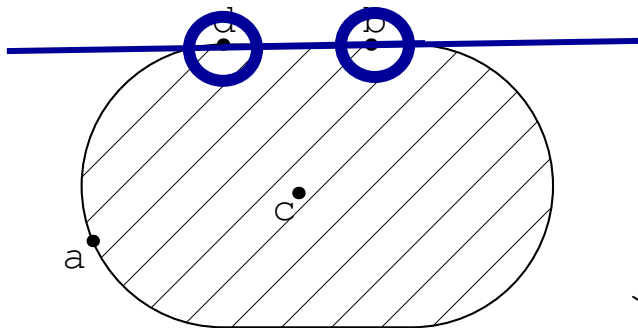
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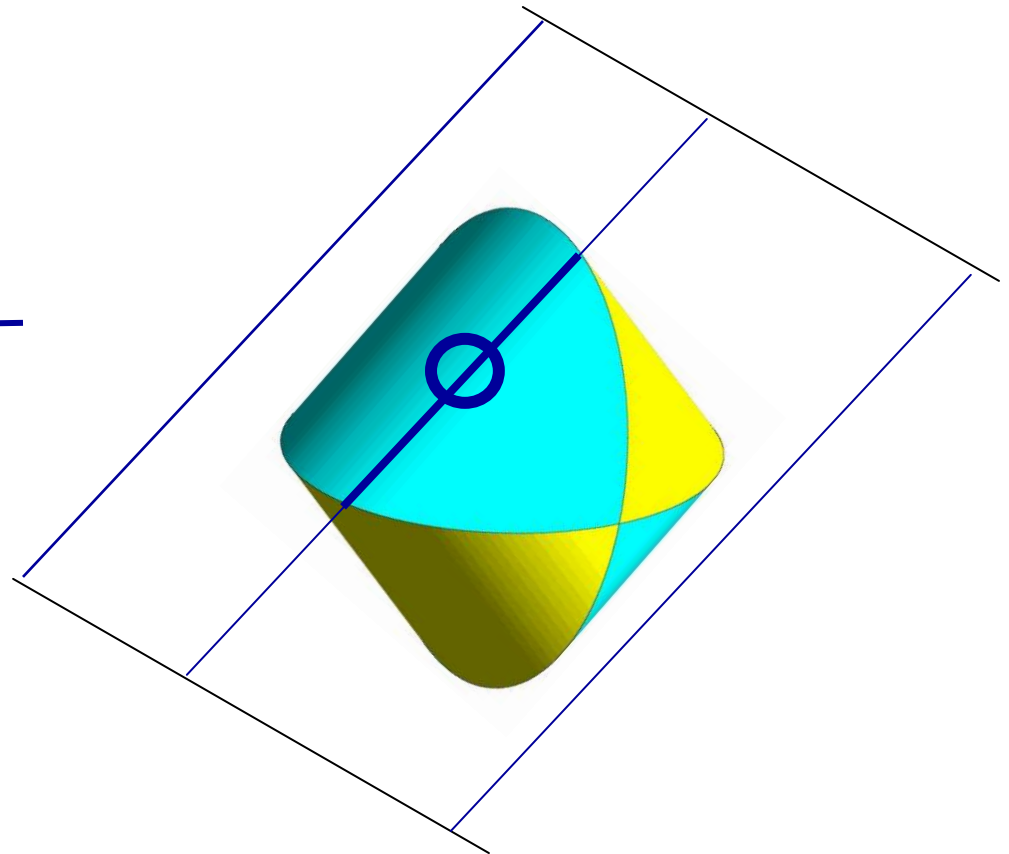
1-exposed

Convex exposedness

- k -exposed: exists halfspace with k dim intersection

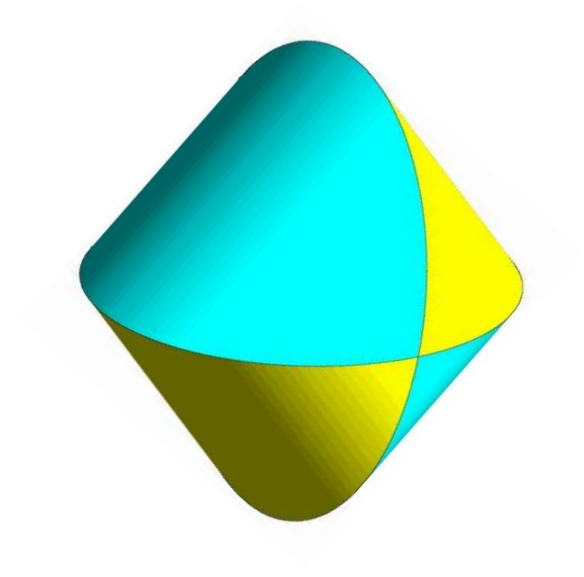
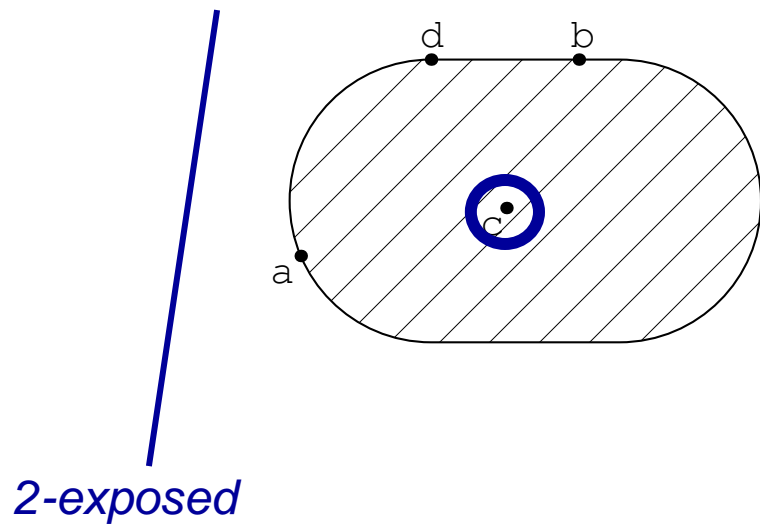


1-exposed



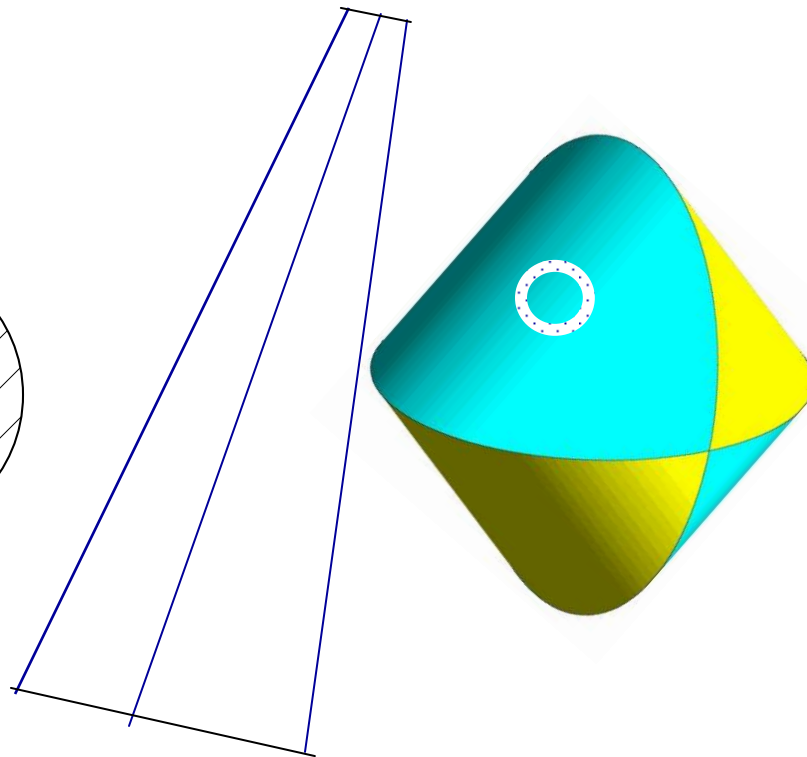
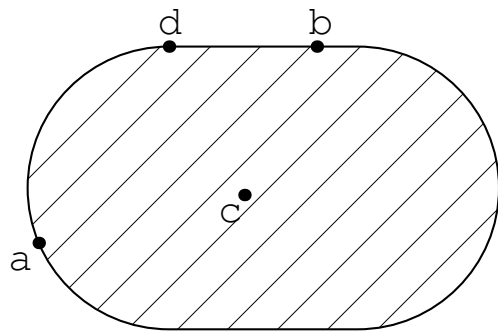
Convex exposedness

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Convex exposedness

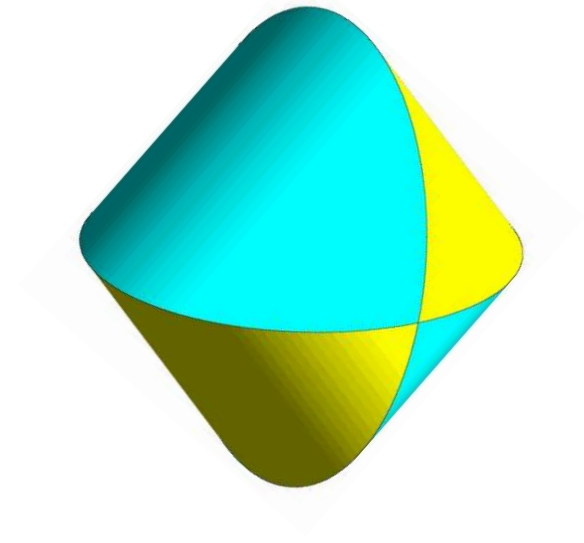
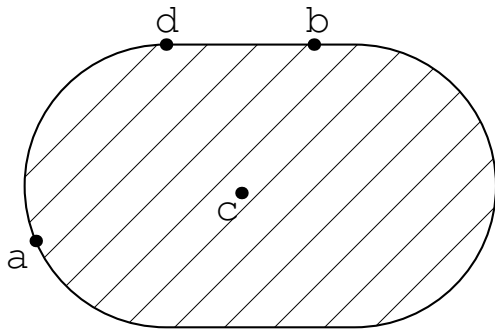
- k -exposed: exists halfspace with k dim intersection



3-exposed

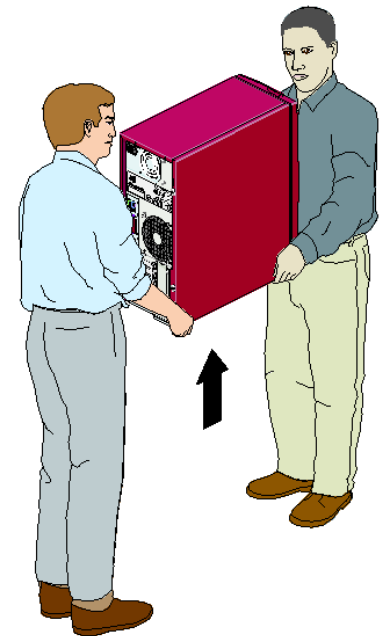
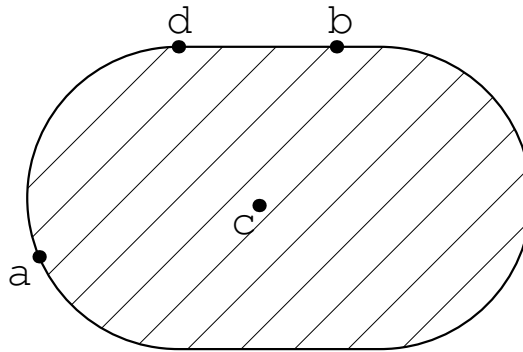
Convex exposedness

- k -exposed implies k -extreme
- But not vice-versa



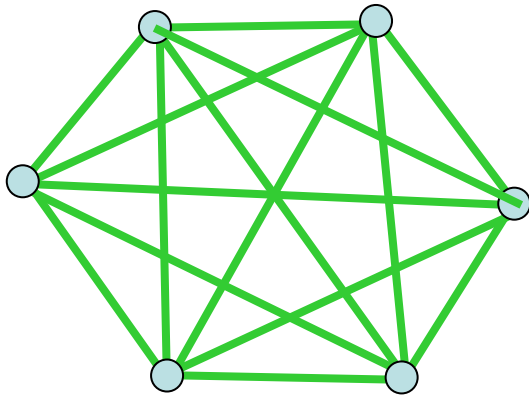
Straszewicz and Asplund

- [Stra 35]: The 0-exposed are dense in the 0-extreme
- [Asp 63]: The k -exposed are dense in the k -extreme



Onto our convex body

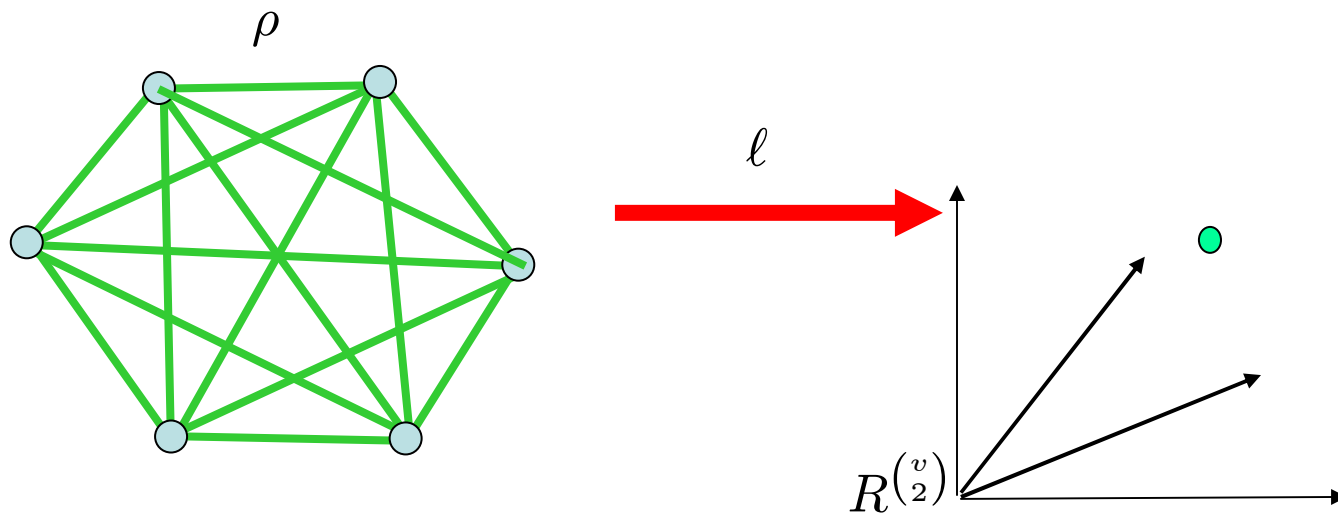
- We will be interested in our graph G
- But we will start with the complete graph on v -verts



Measuring the simplex

Δ_v : the simplex on v vertices

$\ell(\rho) \in R^{\binom{v}{2}}$: squared edge lengths

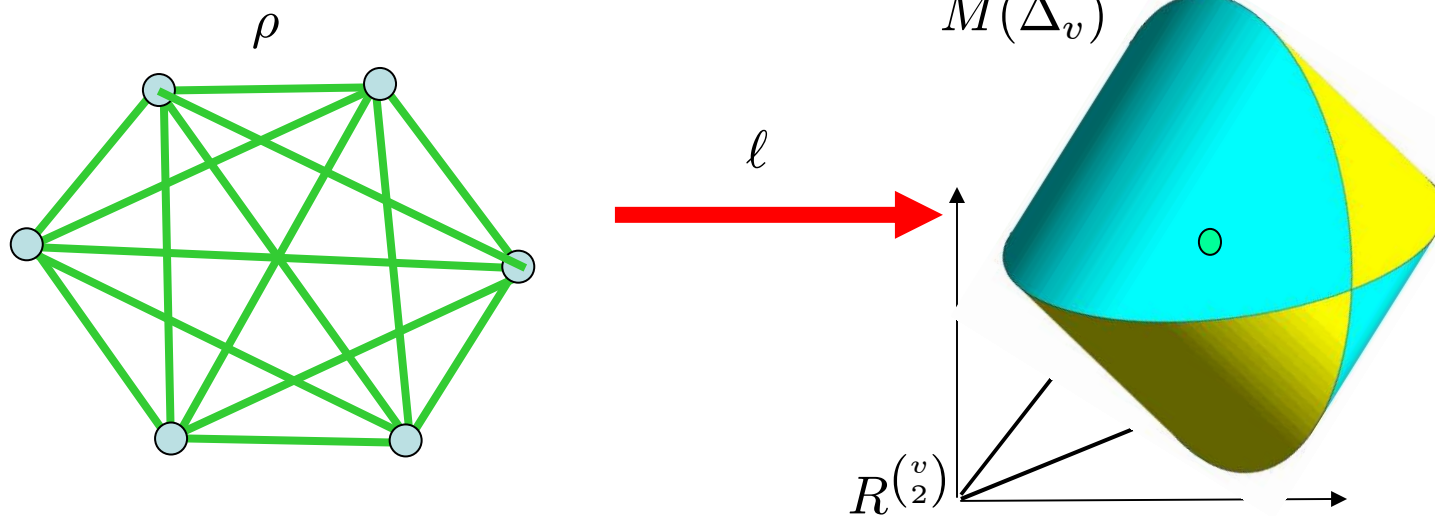


Measuring the simplex

Δ_v : the simplex on v vertices

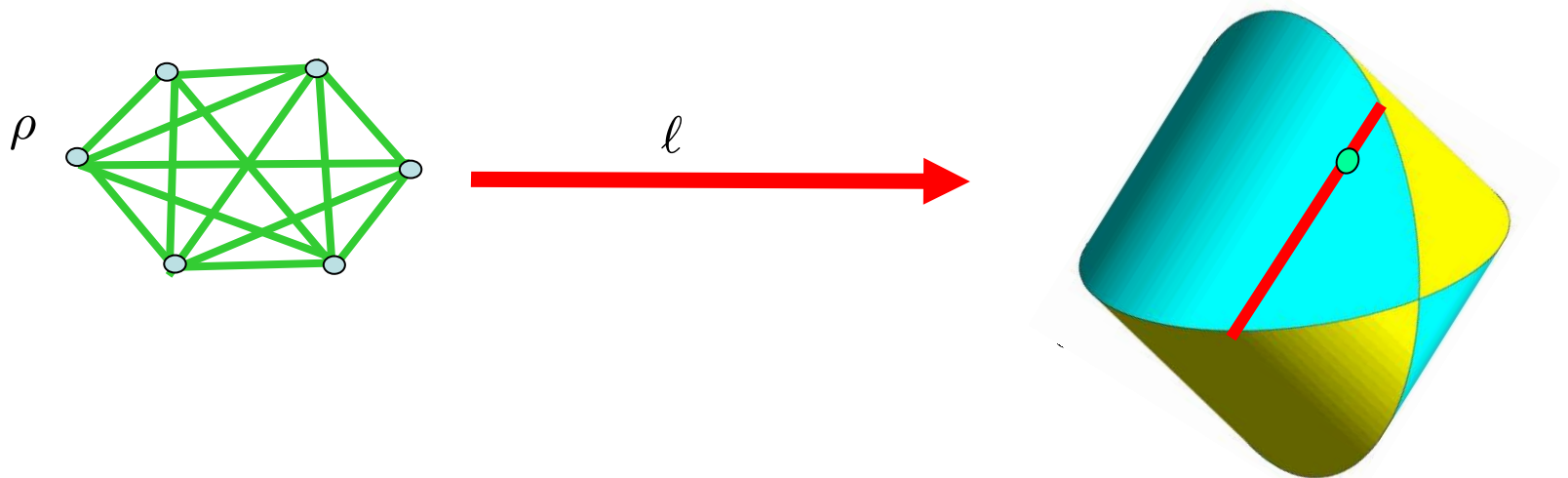
$\ell(\rho) \in R^{\binom{v}{2}}$: squared edge lengths

$M(\Delta_v)$: image of ℓ (any dim span)



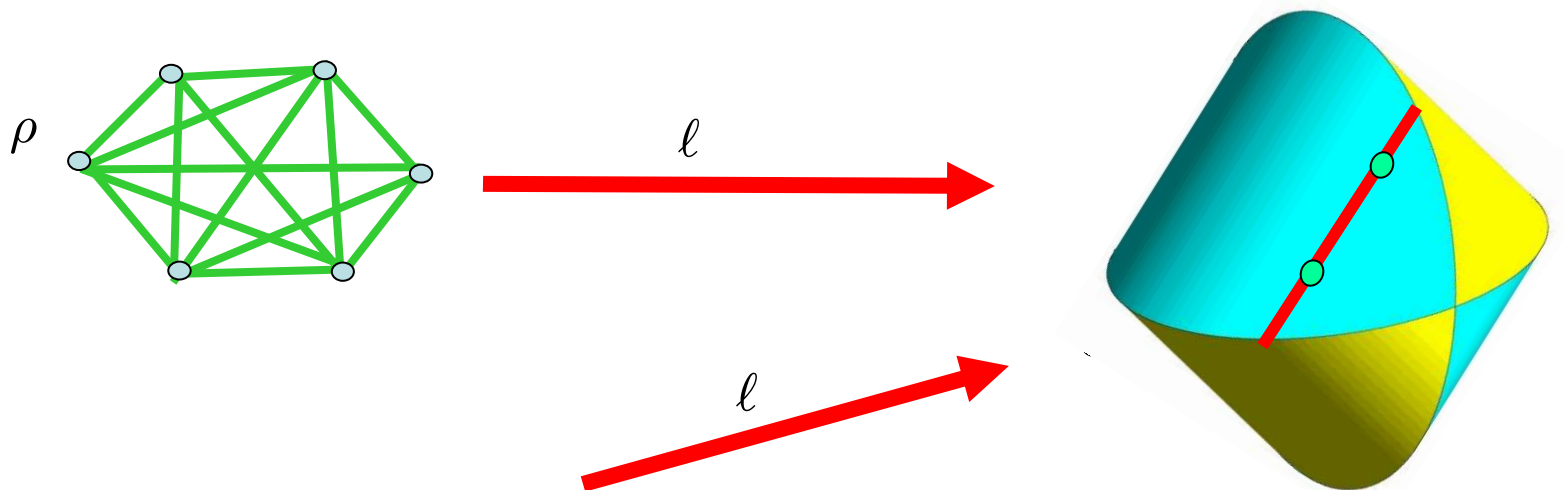
Face structure of $M(\Delta_v)$

$M(\Delta_v)$ is isometric to cone of PSD $(v-1)$ by $(v-1)$ matrices
 $l(q) \in \text{Face}(l(p)) \iff q$ is affine image of p



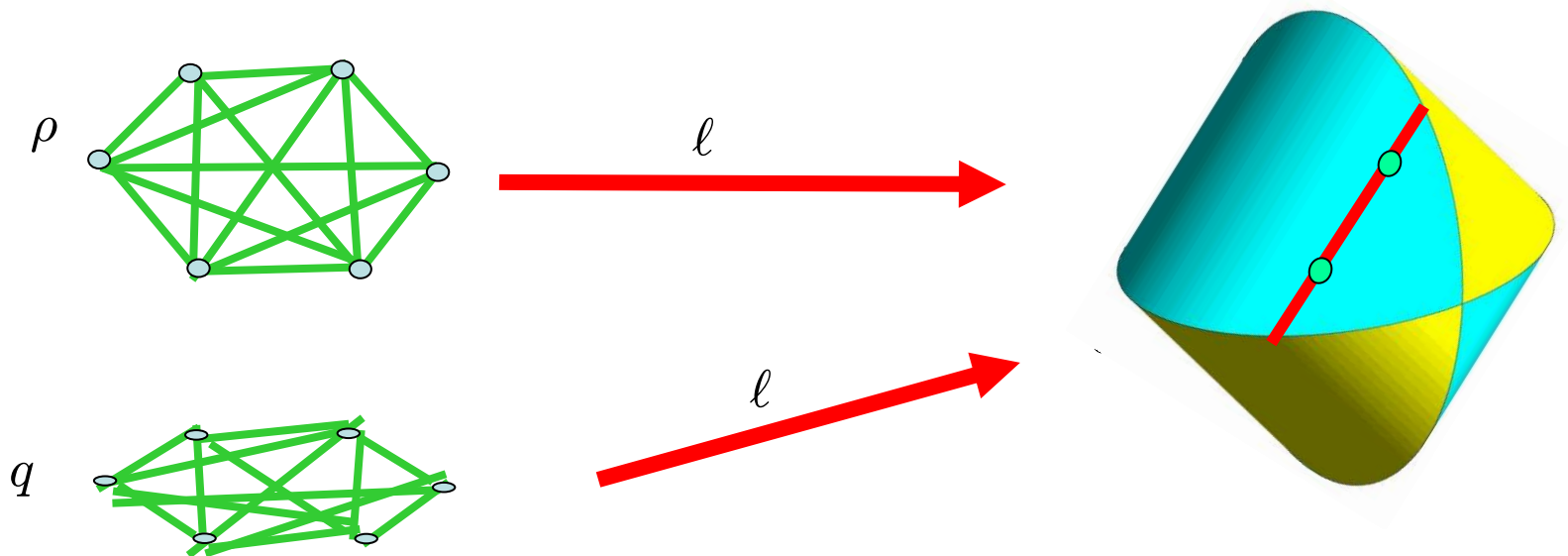
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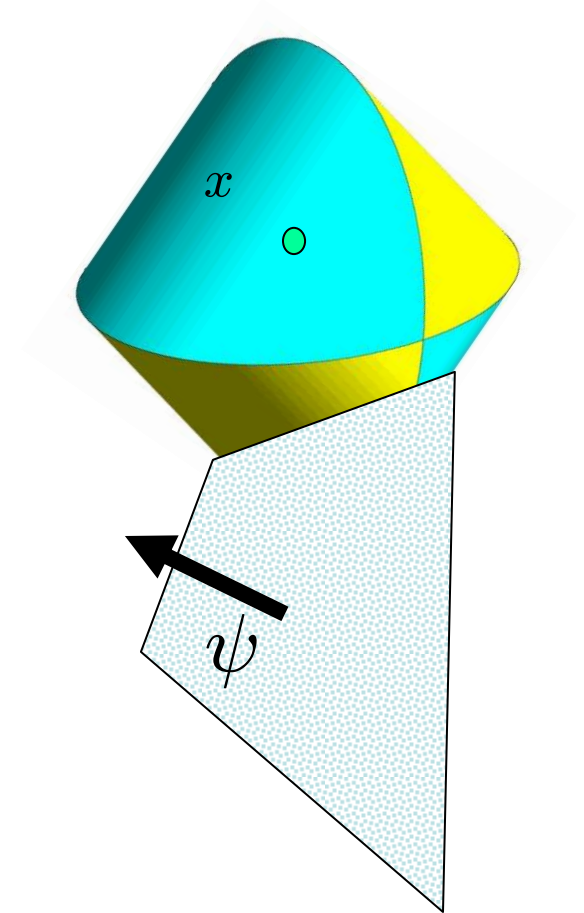
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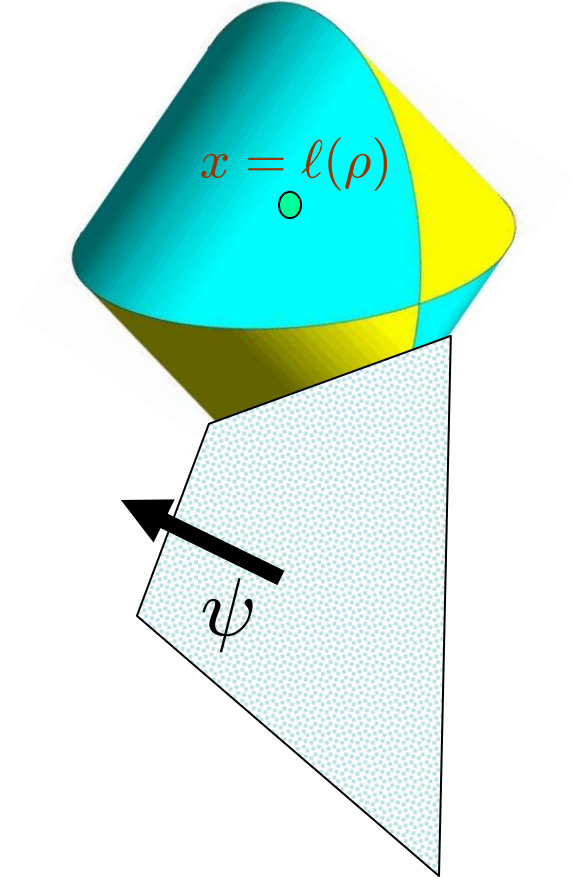
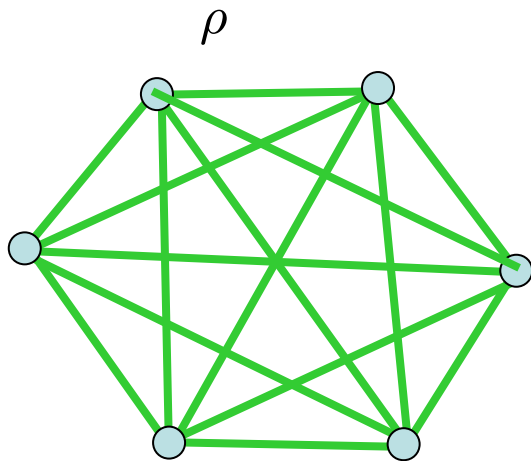
Hyperplane distance

$$x \cdot \psi$$



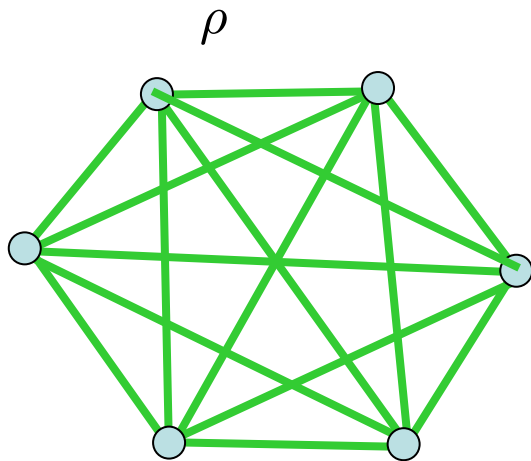
Pull Back

$$x \cdot \psi = \ell(\rho) \cdot \psi$$

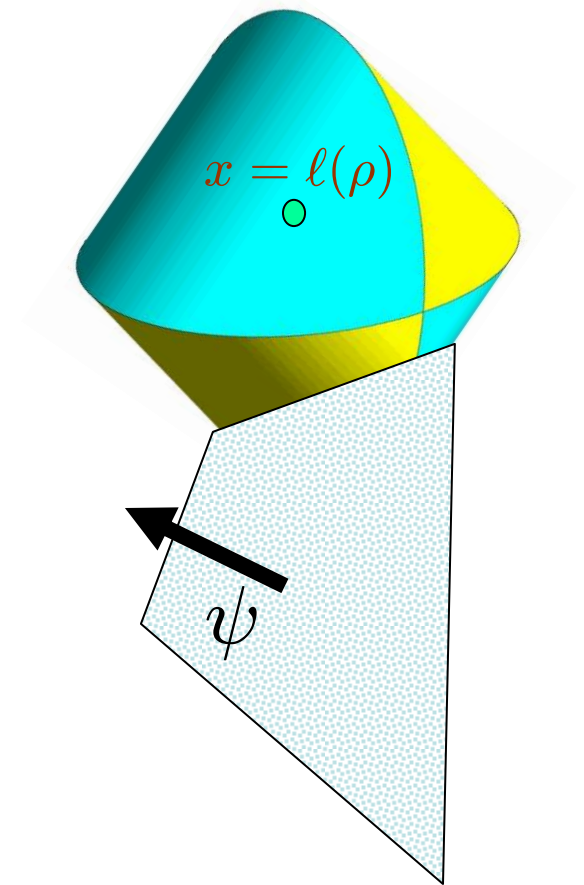


Matricize

$$x \cdot \psi = \ell(\rho) \cdot \psi = \sum_j (\rho^j)^t \Psi \rho^j$$



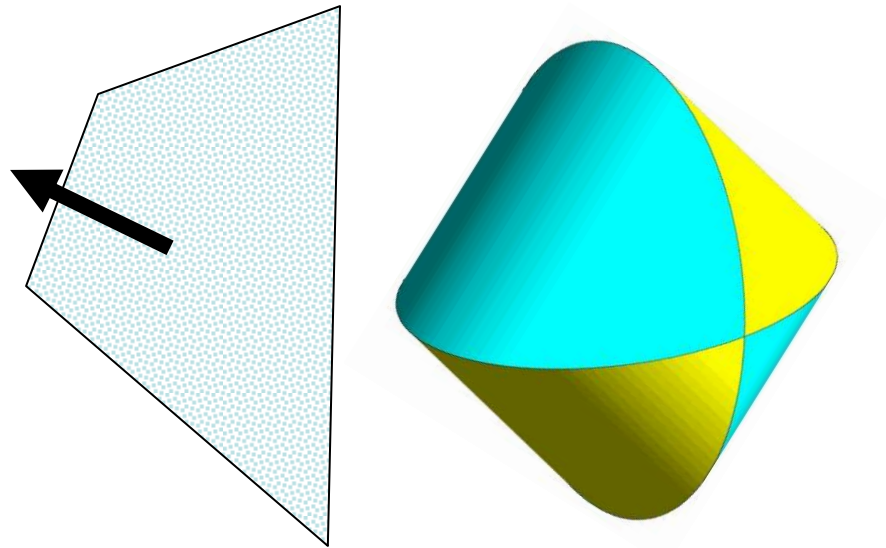
Ψ



Bounding Hyperplane

$$\forall \rho \quad 0 \leq \sum_j (\rho^j)^t \Psi \rho^j$$

Ψ is PSD



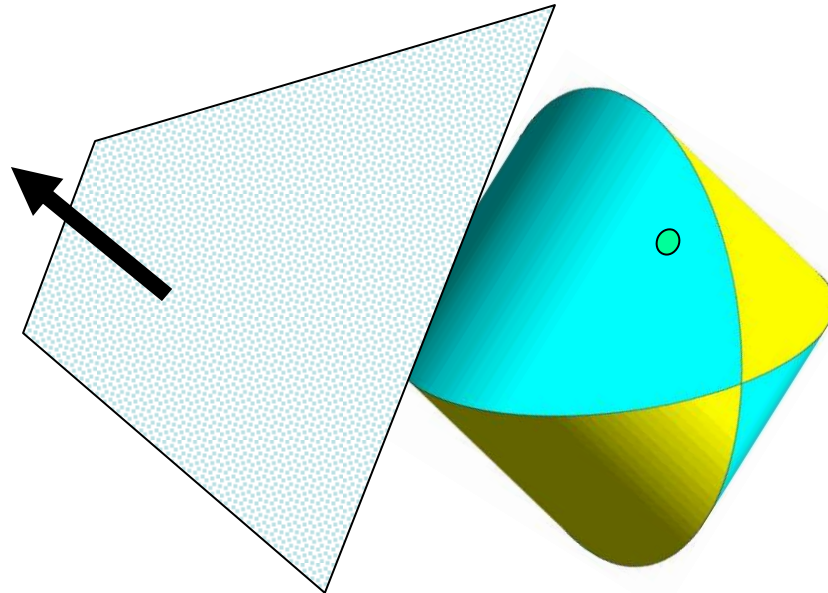
Supporting Hyperplane

Ψ is PSD. $\sum_j (\rho^j)^t \Psi \rho^j = 0$

$\Psi \rho^j = 0$

ρ^j is in the kernel

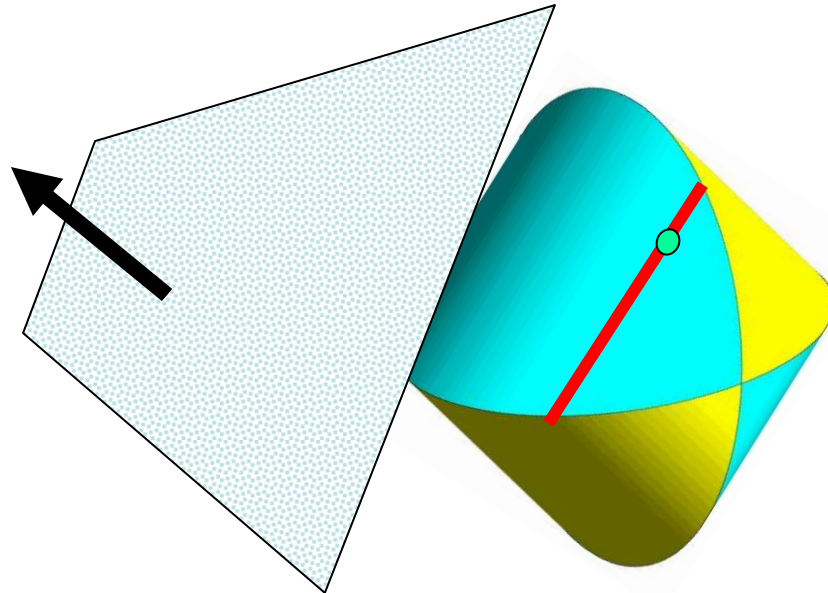
PSD contact == kernel == stress



Supporting Hyperplane

PSD contact at ρ

Contact at the face of $\ell(\rho)$ (at least)

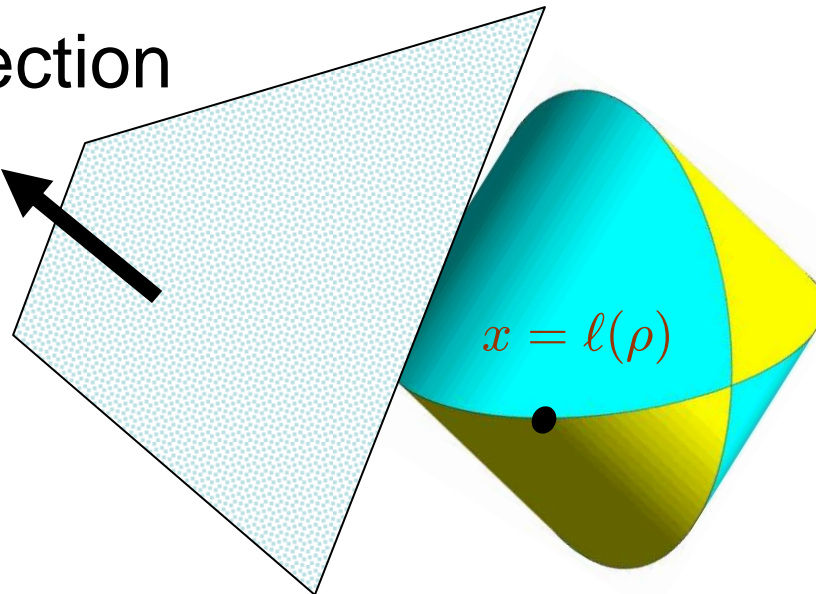


Starting the argument

- We assume that p is generic framework of G and is UR
- We want to find a PSD ESM with rank $v-d-1$

What do we need

- PSD ESM matrix
 - Need a supporting hyperplane of $M(\Delta_v)$
 - With intersection at x
- Rank $v-d-1$
 - **Only** affines of p are in kernel
 - Hyperplane intersection **exactly** Face $\ell(\rho)$



What **else** do we need

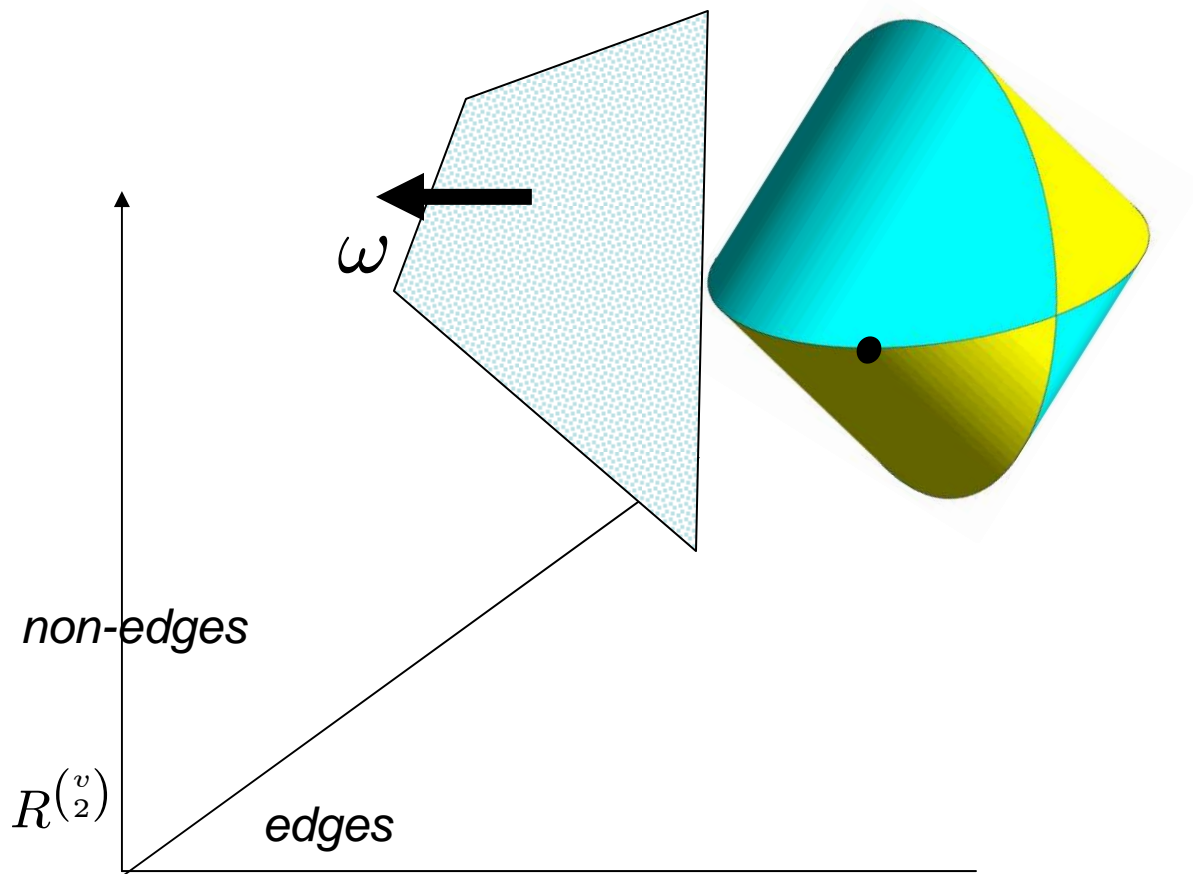
- Recall that we are interested in the graph G , not the simplex

What **else** do we need

- We need zeros at non edges of G

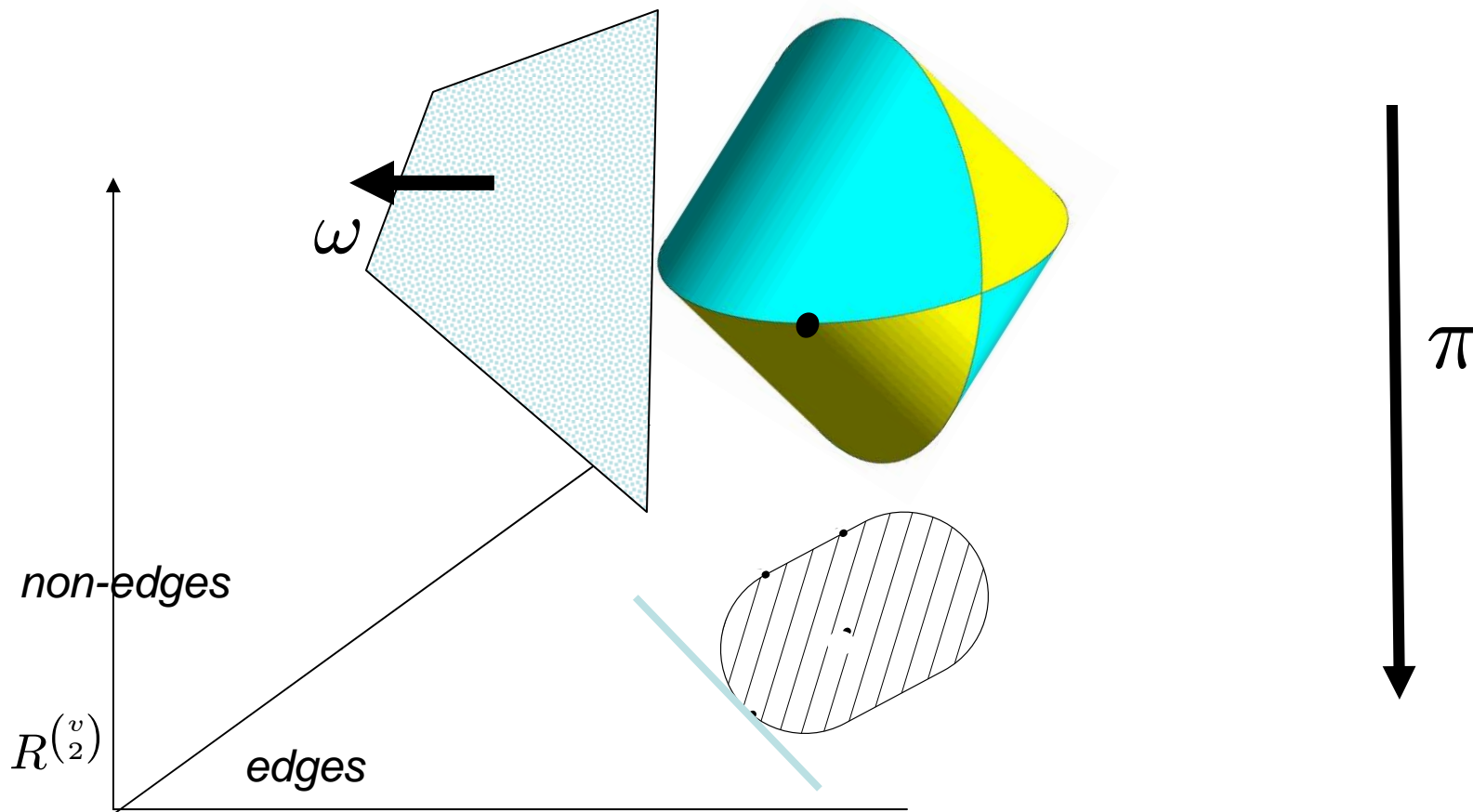
Lets model this

- We need zeros at non edges of G



Lets model this

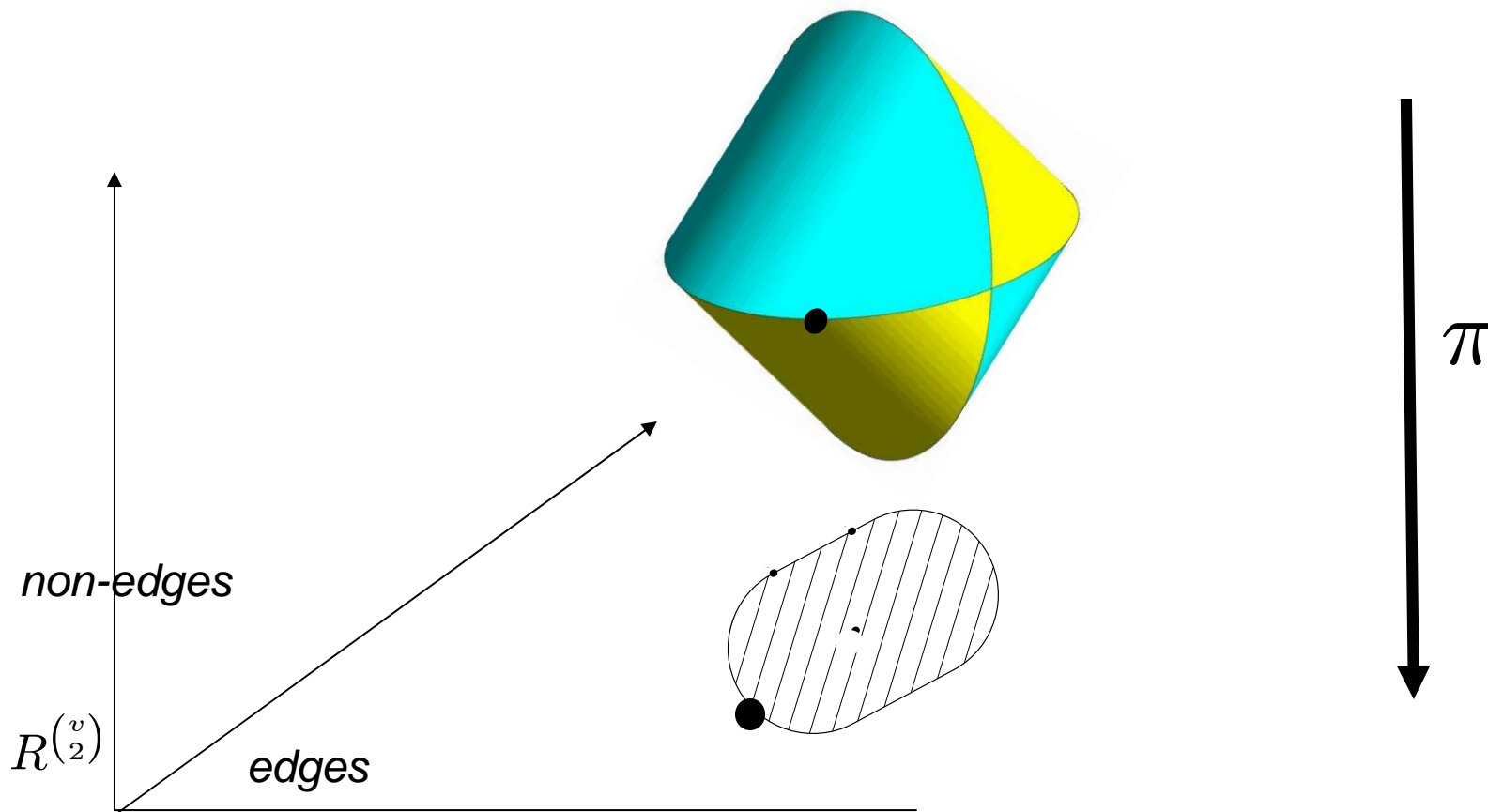
- We need zeros at non edges of G
 - Projects down to a hyperplane



Faces correspond

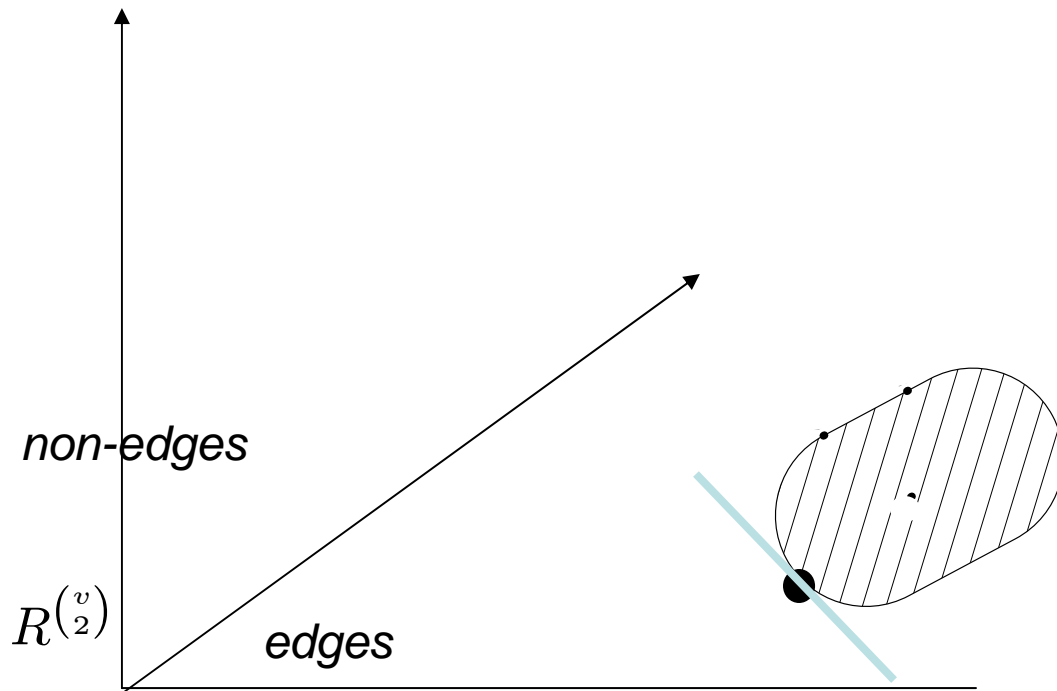
Lemma:

Universal rigidity: $\pi^{-1}(\text{Face}(\pi(x))) = \text{Face}(x)$



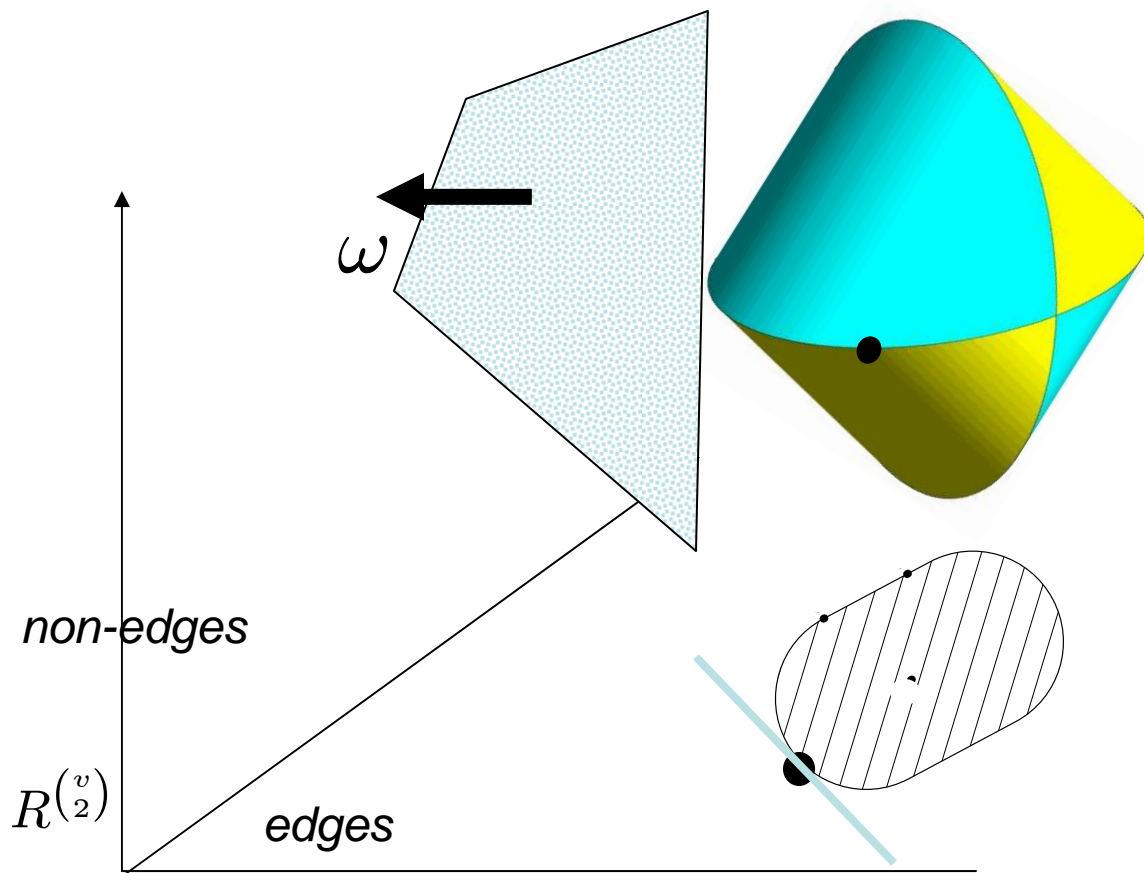
So we really just need

Any hyperplane downstairs with support $\text{Face}(\pi(\ell(\rho)))$



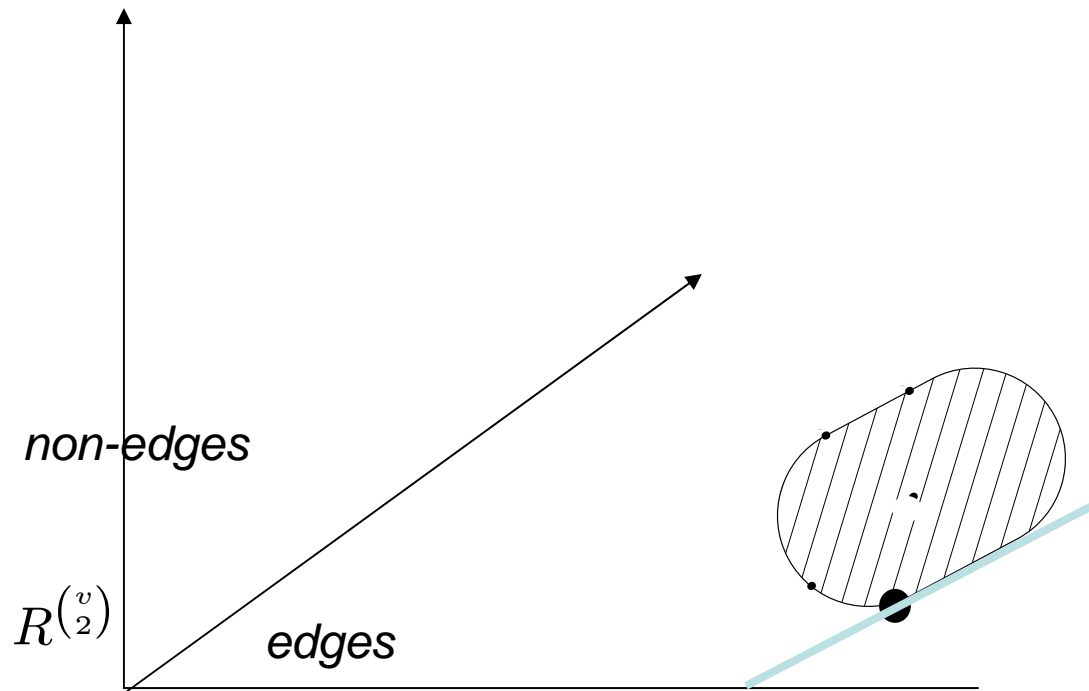
So we really just need

Any hyperplane downstairs with support $\text{Face}(\pi(\ell(\rho)))$
upstairs has support $\text{Face}(\ell(\rho))$



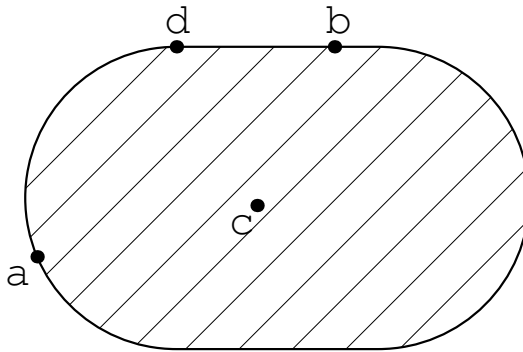
What Could go Wrong

Any hyperplane downstairs with support $\text{Face}(\pi(\ell(\rho)))$



Straszewicz and Asplund

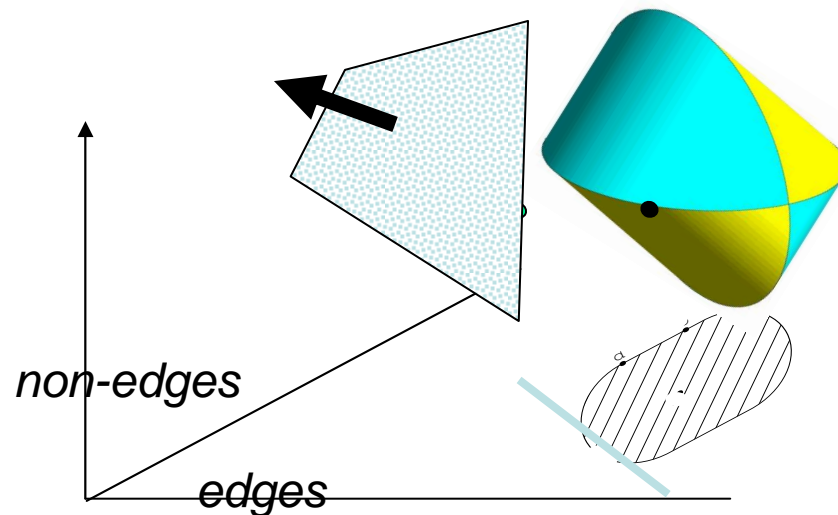
- [Stra 35]: The 0-exposed are dense in the 0-extreme
- [Asp 63]: The k -exposed are dense in the k -extreme



- Use genericity of p (+ a few technicalities)

Now we are done

- Generic point upstairs, maps to generic downstairs
- Downstairs point has supporting plane with right contact (Asplund)
- Corresponds to PSD stress of G
- Contact dim upstairs gives right rank



Computation and SDP




The Harvard Mark I

Gram UR testing

- Suppose generic d -dimensional framework
- Given distances, set up SDP feasibility to look for embeddings
 - Complexity of (exact) SDP feasibility unknown
- UR \Rightarrow only $\text{rank}(x) = d+1$

Lin span = aff span + 1



Gram UR testing

- Suppose generic d -dimensional framework
- Given distances, set up SDP feasibility to look for embeddings
 - Complexity of (exact) SDP feasibility unknown
- UR \Rightarrow only $\text{rank}(x)=d+1$
-details.....
- Not UR \Rightarrow there is x with $\text{rank} > d+1$
- UR: highest rank x is $d+1$

Stress UR testing

- Suppose generic d -dimensional framework
- Given distances, can set up SDP feasibility to look for PSD stresses
- UR: highest rank (Ω) is $v-d-1$ (thm)

Stress UR testing

- Suppose generic d -dimensional framework
- Given distances, can set up SDP feasibility to look for PSD stresses
- UR: highest rank (Ω) is $v-d-1$ (thm)
- So is it any easier??

Dual Computation

- Suppose generic d -dimensional framework
- Given distances, can set up SDP feasibility to look for PSD stresses
 - This is simply “the dual” SDP

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Dual Computation

- Suppose generic d-dimensional framework
- Given distances, can set up SDP feasibility to look for PSD stresses
 - This is simply “the dual” SDP
 - UR \Rightarrow highest rank x is $d+1$
 - UR \Rightarrow highest rank Ω is $v-d-1$ (thm)
 - UR \Rightarrow highest $[\text{rank}(x)+\text{rank}(\Omega)]$ is v (thm)
 - (x, Ω) satisfy strict complementarity if UR (thm)

Strict complementarity

$$\begin{aligned} & \text{Min } \langle x, \beta \rangle \\ & x \in L + b \\ & x \in S_+^n \\ & \textit{primal} \end{aligned}$$

$$\begin{aligned} & \text{Min } \langle b, \Omega \rangle \\ & \Omega \in L^\perp + \beta \\ & \Omega \in S_+^n \\ & \textit{dual} \end{aligned}$$

Strict complementarity

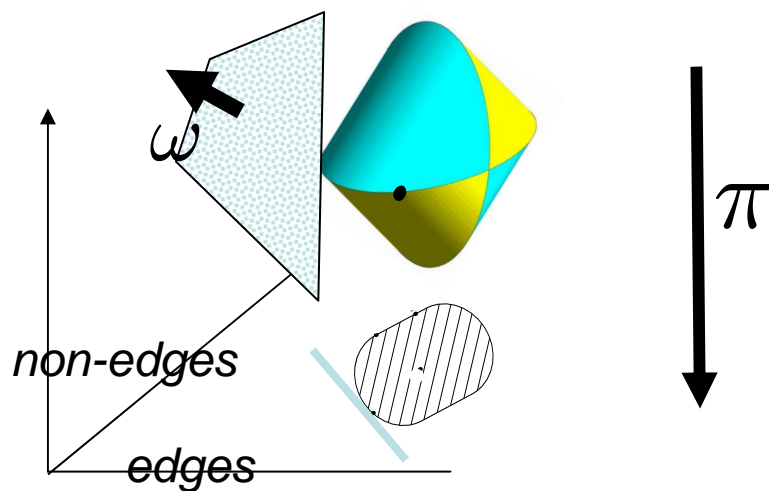
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- WC: $\text{rank}(x^*) + \text{rank}(\Omega^*) \leq n$
 - Holds for all solution pairs
- SC: $(x^*, \Omega^*) \mid \text{rank}(x^*) + \text{rank}(\Omega^*) = n$
 - Strong form of duality
 - Does not always hold for SDP
 - Needed for fast convergence

What we really have proven...

- We did not rely on any properties of the specific projection.



Strict complementarity

$$\begin{aligned} \text{Min } \langle x, \beta \rangle \\ x \in L + b \\ x \in S_+^n \end{aligned}$$

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[GT 09] $\forall(L, b, \beta) \mid x^*$ is unique and generic in its rank over $\mathbb{Q}(L, \beta)$,
 $\exists \Omega^* \mid (x^*, \Omega^*)$ SC

UR

Generic d-dim framework

Compare to [AHO 97]

$$\begin{aligned} &\text{Min } \langle x, \beta \rangle \\ &x \in L + b \\ &x \in S_+^n \end{aligned}$$

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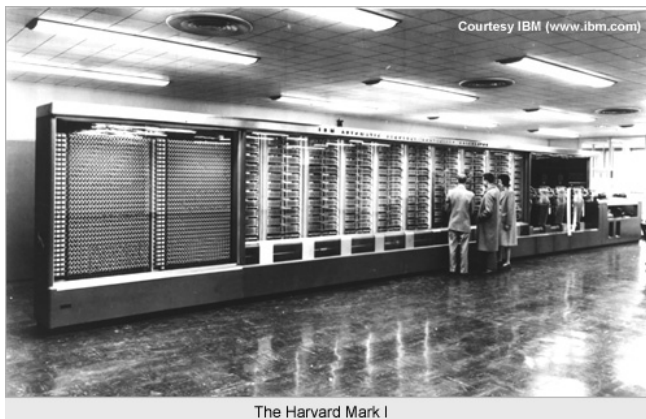
[AHO 97] $\forall L$ over \mathbb{Q} , \forall generic (b, β) ,
 $\exists (x^*, \Omega^*)$ SC

Random lengths= not low rank



Remaining Questions

- Complexity of UR testing
- Is there a single natural unifying statement for strict complementarity
- Graphs that are always UR
- Graphs that are sometimes UR



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Courtesy IBM (www.ibm.com)

Remaining Questions

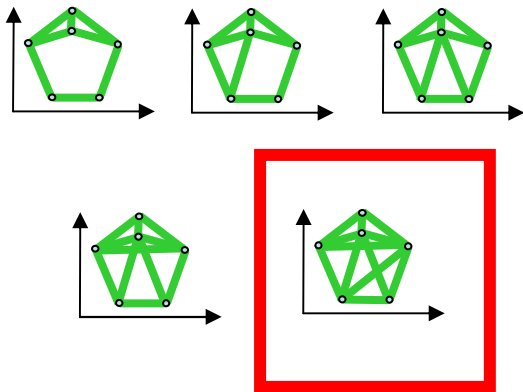
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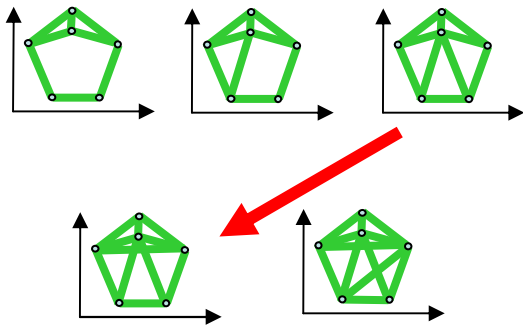
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Thank You