

Problem set 5, ORF523

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1 Problem 1

Notation $E_{ij} = (\delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk})_{k,l}$ the matrix with all 0 except in (i, j) and (j, i)

$$\begin{array}{ll} -\nu(G) = \min_X & Tr(X(-J)) \\ \text{subject to} & X \geq 0 \\ & Tr(XI_n) = 1 \quad (: \alpha) \\ & Tr(E_{ij}X) = 0 \quad \forall (i, j) \in E, i < j \quad (: \lambda_{ij}) \end{array}$$

has for dual:

$$\begin{array}{ll} \max_{\alpha, \lambda_{ij} \in \mathbb{R}} & \alpha \\ \text{subject to} & \alpha I + \sum_{(i,j) \in E} \lambda_{ij} E_{ij} \leq -J \end{array}$$

Or:

$$\begin{array}{ll} \max_{\alpha, \lambda_{ij} \in \mathbb{R}} & \alpha \\ \text{subject to} & \alpha I + \sum_{(i,j) \in E, i < j} \lambda_{ij} E_{ij} \leq -J \end{array}$$

Both are strictly feasible:

- for the primal, take $X = \frac{I_n}{n}$
- For the dual, take $\alpha = -2, \lambda_{ij} = 0$

Which proves that the dual and primal are equal. Taking $\beta = -\alpha$, we can write that:

$$\begin{array}{ll} \nu(G) = \min_{\alpha, \lambda_{ij} \in \mathbb{R}} & \beta \\ \text{subject to} & -\beta I + \sum_{(i,j) \in E} \lambda_{ij} E_{ij} \leq -J \end{array}$$

Note that the $(1, 1)$ entry of $-\beta I + \sum_{(i,j) \in E} \lambda_{ij} E_{ij} + J$: $1 - \beta$ should be negative, so we can amend to the constraints that $\beta \geq 1$

$$\begin{aligned}
-\beta I + \sum_{(i,j) \in E} \lambda_{ij} E_{ij} \leq -J &\iff \beta(I - \sum_{(i,j) \in E} \frac{\lambda_{ij}}{\beta} E_{ij}) \geq J \\
&\iff I - \sum_{(i,j) \in E} \frac{\lambda_{ij}}{\beta} E_{ij} \geq \frac{1}{\beta} \mathbf{1}\mathbf{1}^T \\
&\iff \begin{pmatrix} I - \sum_{(i,j) \in E} \frac{\lambda_{ij}}{\beta} E_{ij} & \begin{matrix} 1 \\ \vdots \\ 1 \end{matrix} \\ 1 & \dots & 1 & \beta \end{pmatrix} \geq 0 \quad (\text{By Schur Lemma bc } \beta > 0)
\end{aligned}$$

Let's note this big matrix Z . It is clear that a matrix $Z \in S^{(n+1) \otimes (n+1)}$ is of this form iff it verifies the constraints of the following optimization problem:

$$\begin{array}{ll}
\min_{\alpha, \lambda_{ij} \in \mathbb{R}} & Z_{n+1, n+1} \\
\text{subject to} & Z \geq 0 \\
& Z_{i, n+1} = Z_{ii} = 0 \\
& Z_{i,j} = 0 \forall \{i, j\} \in \bar{E}
\end{array}$$

Which is then equal to $\vartheta(G)$

Let $C = \chi(\bar{G})$. By definition, there exist a partition of V : $\{V_1, \dots, V_C\}$ such that V_i is a clique for all $i \leq C$

- Define $1_{V_i} \in \mathbb{R}^n$ to be the indicator function of the set V_i , and note that $1 = \sum_{i \leq C} V_i$
- Define $z_i = \begin{pmatrix} 1_{V_i} \\ 1 \end{pmatrix} \in \mathbb{R}^{n+1}$. Note that:

$$z_i z_i^T = \begin{pmatrix} 1_{V_i} 1_{V_i}^T & 1_{V_i} \\ 1_{V_i}^T & 1 \end{pmatrix}$$

- Define

$$Z = \sum_i z_i z_i^T = \begin{pmatrix} \sum 1_{V_i} 1_{V_i}^T & 1 \\ 1^T & C \end{pmatrix}$$

. Z is positive semidefinite because it is a sum of psd terms $z_i z_i^T$

- $(1_{V_i} 1_{V_i}^T)_{kl} = (e_k^T 1_{V_i})(e_l^T 1_{V_i}) = 1_{V_i}(k) 1_{V_i}(l)$. If $(k, l) \in \bar{E}$, then the k^{th} node and the l^{th} node cannot be in the same V_i , and therefore $(1_{V_i} 1_{V_i}^T)_{kl} = 0$
- If $k = l$, all the terms in $\sum_i (1_{V_i} 1_{V_i}^T)_{kl}$ are zero except for the i for which the k^{th} node is in V_i , in which case it is equal to one.

As a conclusion, Z verifies all constraints of the dual, and $Z_{n+1, n+1} = C = \chi(\bar{G})$, so

$$\chi(\bar{G}) \geq \vartheta(G)$$

2

Consider $G = C_5$.

Using CVX to calculate $\vartheta(G)$

```

1 n = 5
2 J = ones(n, n);
3 cvx_begin sdg
4 variable X(n, n) symmetric;
5 maximize(trace(X*J))
6 X >= 0
7 X(5, 1) == 0
8 for i=1:4
9     X(i, i+1) == 0
10 end
11 trace(X) == 1
12 cvx_end
13 ans=cvx_optval

```

2.2361

$$2 < \vartheta(G) < 3$$

- $\vartheta(G) \notin \mathbb{N}$
- $\alpha(G), \chi(\bar{G}) \in \mathbb{N}$

No inequality can thus be tight.

2 Q2

1 —————Duality—————
 Let $x \in [0, 1]^n$ such that C_1, \dots, C_k
 Consider

$$Y = \begin{pmatrix} x_1 & & & x_1 \\ & x_2 & & 0 & x_2 \\ & & \ddots & & \vdots \\ & & & 0 & x_n & x_n \\ x_1 & x_2 & \dots & x_n & 1 \end{pmatrix}$$

2

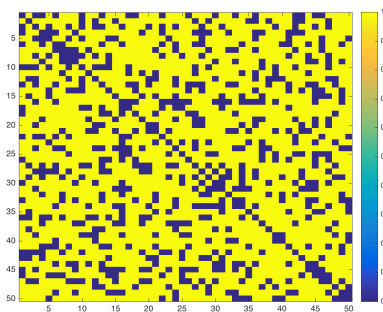


Figure 1: G Adjacency matrix

2.1 $\vartheta(G)$

```

1  n = 50
2  J = ones(n, n);
3
4  cvx_begin sdP
5  variable X(n, n) symmetric;
6  maximize(trace(X*J))
7  X >= 0
8  for i=1:n
9      for j=1:i
10         if G(i, j) == 1
11             X(i, j) == 0
12         end
13     end
14 end
15 trace(X) == 1
16 cvx_end
17
18 ans=cvx_optval

```

5

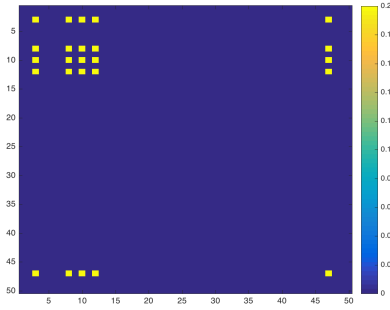


Figure 2: X optimal solution

Note that the resulting X is of rank 1, so it can be decomposed into $X = xx^T$. We check that $V_x = \{i, x_i \neq 0\}$ represents indeed a stable set.

```

1  [v,e] = eigs(full(X),1);
2  stableset = find(abs(v) > 0.01)
3  ans=stableset'

```

3 8 10 12 47

```

1  G(stableset, stableset)

```

Table 1: Subgraph of the nodes in the stableset

0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

Let's assume that there exist another stable set of size 5 V_y .

This would mean that there exist $v \in V_x$ such that imposing $X_{jj} = 0$ would not change α . Let's check:

```

1  n = 50
2  J = ones(n, n);
3  opt = [stableset, zeros(5, 1)]
4  for vi=1:5
5      v = stableset(vi)
6      cvx_begin sdp
7      variable Y(n, n) symmetric;
8      variable optvalue;
9      maximize(trace(Y*J))
10     Y >= 0
11     for i=1:n
12         for j=1:i
13             if G(i, j) == 1
14                 Y(i, j) == 0
15             end
16         end
17     end
18     Y(v,v) == 0
19     trace(Y) == 1
20     optvalue == trace(Y*J)
21     cvx_end
22     opt(vi, 2) = optvalue
23 end
24 ans=opt

```

Table 2: Lovasz	
Node removed	Lovasz of the subgraph
3	4.4463
8	4.5191
10	4.512
12	4.5586
47	4.4771

Since Lovasz number ϑ is an upper bound on α , This proves that any stable set not containing one of the nodes in V_x is of size less than 5.

We have just proved uniqueness of the stable set.

2.2 μ^{LP}

$k = 2$

```

1 cvx_begin
2 variable x(n)
3 maximize(sum(x))
4 for i=2:n
5     for j=1:(i-1)
6         if G(i, j) == 1
7             x(i) + x(j) <= 1
8         end
9     end
10 end
11 0 <= x <= 1
12 cvx_end
13
14 ans=cvx_optval

```

25

k = 3

```

1 cvx_begin
2 variable x(n)
3 maximize(sum(x))
4 for i=2:n
5     for j=1:(i-1)
6         if G(i, j) == 1
7             x(i) + x(j) <= 1
8         end
9         for r=1:(j-1)
10            if G(i, j) + G(j, r) + G(r, i) == 3
11                x(i) + x(j) + x(r) <= 1
12            end
13        end
14    end
15 end
16 0 <= x <= 1
17 cvx_end
18
19 ans=cvx_optval

```

16.667

k = 4

```

1 M = 50
2 cvx_begin
3     variable x(n)
4     maximize(sum(x))
5     for i=2:M
6         for j=1:(i-1)

```

```

7         if G(i, j) == 0
8             continue
9         end
10        x(i) + x(j) <= 1
11        for r=1:(j-1)
12            if G(j, r) == 0 || G(r, i) == 0
13                continue
14            end
15            x(i) + x(j) + x(r) <= 1
16            for p =1:(r-1)
17                if G(i, p) == 0 || G(j, p) == 0 || G(r, p) == 0
18                    continue
19                end
20                x(i) + x(j) + x(r) + x(p) <= 1
21            end
22        end
23    end
24 end
25 0 <= x <= 1
26 cvx_end
27
28 ans=cvx_optval

```

12.5

3 Problem 3

1. Let $(a, b), (u, v)$ be two nodes in $G_A \otimes G_B$ The two nodes are connected if and only if:

- $A_{au} = 1, A_{bv} = 1$
- $a = u, A_{bv} = 1$
- $A_{au} = 1, b = v$

This can be summerised as $(A_{au} + \delta_{au})(A_{bv} + \delta_{bv}) - \delta_{au}\delta_{bv} = 1$

So the adjacency matrix of $G_A \otimes G_B$ is $(A + I_n) \otimes (B + I_m) - I_{nm}$.

Where \otimes denote the Kronecker product: $(A \otimes B)_{p(r-1)+v, q(s-1)+w} = A_{rs}B_{vw}$

2.

4 Problem 4

1. (1) is equivalent to

$$\begin{cases} x^T A y &= \max_{\tilde{x} \in \Delta_m} \tilde{x}^T A y \\ x^T B y &= \max_{\tilde{y} \in \Delta_n} x^T B \tilde{y} \end{cases}$$

Consider the first problem:

$$\max_{\tilde{x} \in \Delta_m} \tilde{x}^T A y$$

This is an LP whose feasible region $\Delta_m = \text{conv}(e_i, i = 1 \dots m)$ is compact, so the maximum is attained in one of the extreme points e_{i_0} . Therefore

$$x^T Ay = \max_{\tilde{x} \in \Delta_m} \tilde{x}^T Ay \iff x^T Ay = e_{i_0}^T Ay = \max_i e_i^T Ay \iff x^T Ay \geq e_i^T Ay \forall i$$

Same argument applies for y so that:

$$x^T By = \max_{\tilde{y} \in \Delta_n} x^T B\tilde{y} \iff x^T Ay \geq x^T Be_i \forall i$$

So:

$$(1) \iff \begin{cases} x^T Ay \geq e_i^T Ay \forall i = 1 \dots m \\ x^T By \geq x^T Ae_i \forall i = 1 \dots n \end{cases}$$

2.

$$x \in \Delta_m, y \in \Delta_n$$

$$\text{Note } z = \begin{pmatrix} x \\ y \end{pmatrix}, u = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, v = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, M = zz^T = \begin{pmatrix} xx^T & xy^T \\ yx^T & yy^T \end{pmatrix}$$

Note that

- $M \succeq 0$
- $\text{rank}(M) = 1$
- $u^T Mu = 1^T xx^T 1 = (1^T x)^2 = 1$, similarly $v^T Mv = 1$
- $M_{(i+m),j} = x_i y_j \geq 0$

Now let $M \in S^{n+m}$, verifying all the previous conditions. Then by cholesky, there exists a vector $z \in \mathbb{R}^{n+m}$, such that: $M = zz^T$

- Let's decompose $z := \begin{pmatrix} x \\ y \end{pmatrix}$
- $1 = u^T Mu \implies \sum_i x_i = \pm 1$, similarly, $\sum y_j = \pm 1$
- $M_{(i+m),j} = x_i y_j \geq 0 \implies x_i, y_j$ all share the same sign.
- We can always change x and/or y to $-x, -y$ to make $x, y \geq 0$ and therefore $\sum_i x_i = \sum_j y_j = 1$, e.g $x \in \Delta_m, y \in \Delta_n$
- $Mu = \begin{pmatrix} xx^T 1 \\ yx^T 1 \end{pmatrix} = \begin{pmatrix} (x^T 1)x \\ (x^T 1)y \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix}$
- $x = \underbrace{\begin{pmatrix} I_n & 0 \\ 0 & 0 \end{pmatrix}}_{J_1} Mu$
- $y = \underbrace{\begin{pmatrix} 0 & 0 \\ 0 & I_m \end{pmatrix}}_{J_2} Mu$
- $yx^T = M_{n:n+m, 1:n}$

The constraint of (2) can be formulated as follow:

•

$$x^T Ay = \text{tr}(yx^T A) = \text{tr}(M_{n:n+m, 1:n} A)$$

- $e_i^T Ay = e_i^T AJ_2Mu = \text{tr}(ue_i^T AJ_2M)$
- $x^T By = \text{tr}(M_{n:n+m,1:n}B)$
- $x^T Be_i = (J_1Mu)^T Be_i = u^T MJ_1Be_i = \text{tr}(MJ_1Be_iu^T)$

In conclusion:

- $M \succeq 0$
- $\text{rank}(M) = 1$
- $M_{i+m,j} \geq 0$
- $\text{tr}(Mu u^T) = 1$
- $\text{tr}(M_{n:n+m,1:n}A) \geq \text{tr}(Mue_i^T AJ_2)$
- $\text{tr}(M_{n:n+m,1:n}B) \geq \text{tr}(MJ_1Be_iu^T)$