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# HW2 ECE 719 Optimal systems

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## 0.1 Problem 1

### PROBLEM DESCRIPTION

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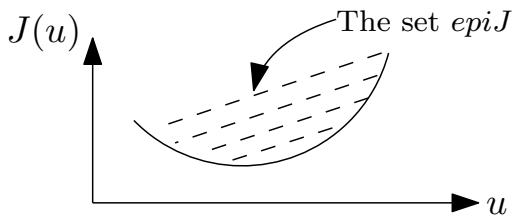
#### ECE 719 – Homework Epigraph

Give a function  $J : \mathbf{R}^n \rightarrow \mathbf{R}$ , we recall that its *epigraph* is the a set in  $\mathbf{R}^{n+1}$  given by

$$\text{epi } J = \{(u, \alpha) \in \mathbf{R}^{n+1} : \alpha \geq J(u)\}.$$

Now prove that  $J$  is a convex function if and only if  $\text{epi } J$  is a convex set.

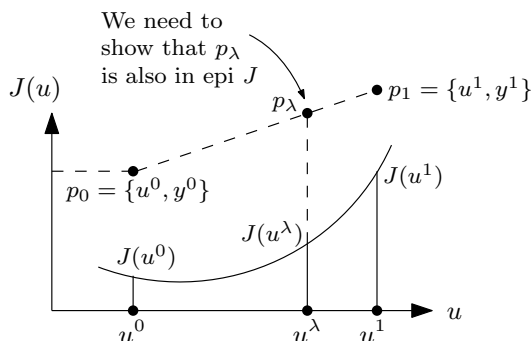
**SOLUTION** The following diagram illustrates  $\text{epi } J$  for  $n = 1$ . In words, it is the set of all points above the curve of the function  $J(u)$



This is an iff proof, hence we need to show the following

1. Given  $J$  is convex function, then show that  $\text{epi } J$  is a convex set.
2. Given that  $\text{epi } J$  is a convex set, then show that  $J$  is a convex function.

Proof of first direction We pick any two arbitrary points in  $\text{epi } J$ , such as  $p_0 = (u^0, y^0)$  and  $p_1 = (u^1, y^1)$ . To show  $\text{epi } J$  is a convex set, we need now to show that any point on the line between  $p_0, p_1$  is also in  $\text{epi } J$ . The point between them is given by  $p_\lambda = (u^\lambda, y^\lambda)$  where  $\lambda \in [0, 1]$ . The following diagram helps illustrates this for  $n = 1$ .



The point  $p_\lambda$  is given by

$$\begin{aligned} (u^\lambda, y^\lambda) &= (1 - \lambda) p_0 + \lambda p_1 \\ &= (1 - \lambda) (u^0, y^0) + \lambda (u^1, y^1) \\ &= ((1 - \lambda) u^0 + \lambda u^1, (1 - \lambda) y^0 + \lambda y^1) \end{aligned}$$

Therefore  $y^\lambda = (1 - \lambda) y^0 + \lambda y^1$ . Since  $p_0, p_1$  are in  $\text{epi } J$ , then by the definition of  $\text{epi } J$ , we know that  $y^0 \geq J(u^0)$  and  $y^1 \geq J(u^1)$ . Therefore we conclude that

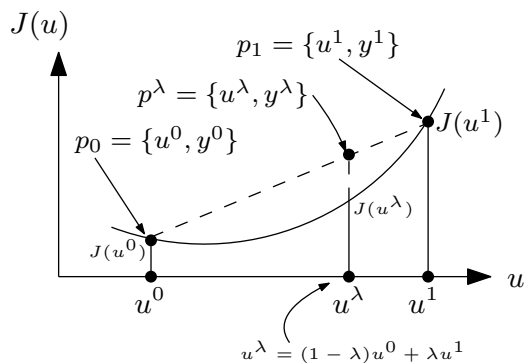
$$y^\lambda \geq (1 - \lambda) J(u^0) + \lambda J(u^1) \quad (1)$$

But since we assumed  $J$  is a convex function, then we also know that  $(1 - \lambda) J(u^0) + \lambda J(u^1) \geq J(u^\lambda)$  where  $u^\lambda = (1 - \lambda) u^0 + \lambda u^1$ . Therefore (1) becomes

$$y^\lambda \geq J(u^\lambda)$$

This implies the arbitrary point  $p_\lambda$  is in  $\text{epi } J$ .

We now need to proof the other direction. Given that  $\text{epi } J$  is a convex set, then show that  $J$  is a convex function. Since  $\text{epi } J$  is a convex set, we pick two arbitrary points in  $\text{epi } J$ , such as  $p_0, p_1$ . We can choose  $p_0 = (u^0, J(u^0))$  and  $p_1 = (u^1, J(u^1))$ . These are still in  $\text{epi } J$ , but on the lower bound, on the edge with  $J(u)$  curve.



Since  $p_0, p_1$  are two points in a convex set, then any point  $p^\lambda$  on a line between them is also in  $\text{epi } J$  (by definition of a convex set). And since  $p^\lambda = (1 - \lambda)p_0 + \lambda p_1$  this implies

$$\begin{aligned}
 p^\lambda &= (u^\lambda, y^\lambda) \\
 &= ((1 - \lambda)p_0 + \lambda p_1) \\
 &= ((1 - \lambda)(u^0, J(u^0)) + \lambda(u^1, J(u^1))) \\
 &= \left( (1 - \lambda)u^0 + \lambda u^1, \overbrace{(1 - \lambda)J(u^0) + J(u^1)}^{y^\lambda} \right)
 \end{aligned} \tag{1}$$

Since  $p^\lambda$  is in  $\text{epi } J$  then by definition of  $\text{epi } J$

$$y^\lambda \geq J(u^\lambda) \tag{2}$$

But from (1) we see that  $y^\lambda = (1 - \lambda)J(u^0) + J(u^1)$ , therefore (2) is the same as writing

$$(1 - \lambda)J(u^0) + J(u^1) \geq J(u^\lambda) \tag{3}$$

But  $u^\lambda = (1 - \lambda)u^0 + \lambda u^1$ , hence the above becomes

$$(1 - \lambda)J(u^0) + J(u^1) \geq J((1 - \lambda)u^0 + \lambda u^1)$$

But the above is the definition of a convex function. Therefore  $J(u)$  is a convex function. QED.

## 0.2 Problem 2

### PROBLEM DESCRIPTION

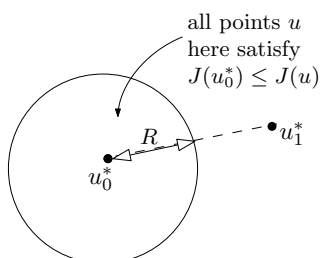
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### ECE 719 – Homework Unique Minimum

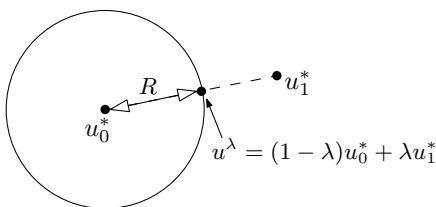
Suppose  $J : \mathbf{R}^n \rightarrow \mathbf{R}$  is strictly convex. Then prove the following: If a minimizing element  $u^* \in \mathbf{R}^n$  exists, it must be unique.

**SOLUTION** Let  $u_0^*$  and  $u_1^*$  be any two different minimizing elements in  $\mathfrak{R}^n$  such that  $J(u_0^*) < J(u_1^*)$ . We will show that this leads to contradiction. Since  $u_0^*$  is a minimizer, then there exists some  $R > 0$ , such that all points  $u$  that satisfy  $\|u^* - u\| \leq R$  also satisfy

$$J(u_0^*) \leq J(u)$$



We will consider all points along the line joining  $u_0^*, u_1^*$ , and pick one point  $u^\lambda$  that satisfies  $\|u^* - u^\lambda\| \leq R$ , where  $\lambda \in [0, 1]$  is selected to make the convex mixture  $u^\lambda = (1 - \lambda)u_0^* + \lambda u_1^*$  satisfied. Therefore any  $\lambda \leq \frac{R}{\|u_0^* - u_1^*\|}$  will put  $u^\lambda$  inside the sphere of radius  $R$ .



Hence now we can say that

$$J(u_0^*) \leq J(u^\lambda) \tag{1}$$

But given that  $J(u)$  is a strict convex function, then

$$J(u^\lambda) < (1 - \lambda)J(u_0^*) + \lambda J(u_1^*) \tag{2}$$

Since we assumed that  $J(u_0^*) < J(u_1^*)$ , then if we replace  $J(u_1^*)$  by  $J(u_0^*)$  in the RHS of (2), it will change from  $<$  to  $\leq$  resulting in

$$\begin{aligned} J(u^\lambda) &\leq (1 - \lambda)J(u_0^*) + \lambda J(u_0^*) \\ J(u^\lambda) &\leq J(u_0^*) \end{aligned} \tag{3}$$

We see that equations (3) and (1) are a contradiction. Therefore our assumption is wrong and there can not be more than one minimizing element and  $u_0^*$  must be the same as  $u_1^*$ .

### 0.3 Problem 3

#### PROBLEM DESCRIPTION

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#### ECE 719 – Homework Global Minimum

**Preamble:** Suppose  $J : \mathbf{R}^n \rightarrow \mathbf{R}$ . A point  $u^* \in \mathbf{R}^n$  is said to be a *local minimum* of  $J$  if there exists some suitably small  $\delta > 0$  leading to satisfaction of the following condition:

$$J(u^*) \leq J(u)$$

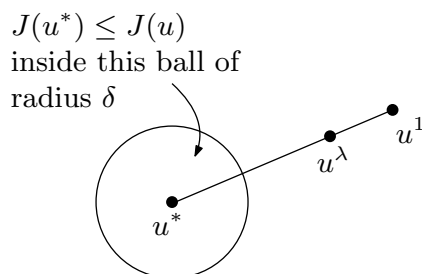
for all  $u$  such that  $\|u - u^*\| < \delta$ . Said another way,  $u^*$  is a minimizing element over a suitably small open neighborhood. For the case when  $J(u^*) \leq J(u)$  for all  $u$ , we call  $u^*$  a *global minimum* of  $J$ .

**The Homework Problem:** Suppose  $J : \mathbf{R}^n \rightarrow \mathbf{R}$  is convex. Prove that every local minimum of  $J$  is a global minimum.

**SOLUTION** We are given that  $J(u^*) \leq J(u)$  for all  $u$  such that  $\|u^* - u\| < \delta$ . Let us pick any arbitrary point  $u^1$ , outside ball of radius  $\delta$ . Then any point on the line between  $u^*$  and  $u^1$  is given by

$$u^\lambda = (1 - \lambda)u^* + \lambda u^1$$

In picture, so far we have this setup

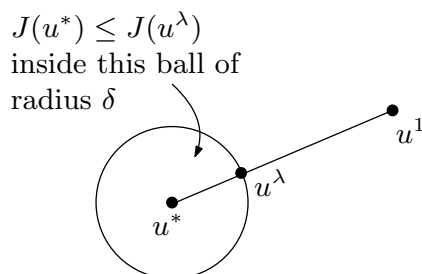


We now need to show that  $J(u^*) \leq J(u^1)$  even though  $u^1$  is outside the ball. Since  $J$  is a convex function, then

$$J(u^\lambda) \leq (1 - \lambda)J(u^*) + \lambda J(u^1) \quad (1)$$

We can now select  $\lambda$  to push  $u^\lambda$  to be inside the ball. We are free to change  $\lambda$  as we want while keeping  $u^1$  fixed, outside the ball. If we do this we then we have

$$J(u^*) \leq J(u^\lambda)$$



Hence (1) becomes

$$J(u^*) \leq (1 - \lambda)J(u^*) + \lambda J(u^1) \quad (2)$$

Where we replaced  $J(u^\lambda)$  by  $J(u^*)$  in (1) and since  $J(u^*) \leq J(u^\lambda)$  the  $\leq$  relation remained

valid. Simplifying (2) gives

$$J(u^*) \leq J(u^*) - \lambda J(u^*) + \lambda J(u^1)$$
$$\lambda J(u^*) \leq \lambda J(u^1)$$

For non-zero  $\lambda$  this means  $J(u^*) \leq J(u^1)$ . This completes the proof, since  $u^1$  was arbitrary point anywhere. Hence  $u^*$  is global minimum. QED

## 0.4 Problem 4

### PROBLEM DESCRIPTION

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#### ECE 719 – Homework Multiple Combinations

For a convex function  $J : \mathbf{R}^n \rightarrow \mathbf{R}$ , prove the following condition is satisfied: Given any points  $u^1, u^2, \dots, u^N \in \mathbf{R}^n$  and any scalars  $\lambda_1, \lambda_2, \dots, \lambda_N \geq 0$  such that

$$\sum_{i=1}^N \lambda_i = 1,$$

it follows that

$$J\left(\sum_{i=1}^N \lambda_i u^i\right) \leq \sum_{i=1}^N \lambda_i J(u^i).$$

### SOLUTION

We need to show that  $J\left(\sum_{i=1}^N \lambda_i u^i\right) \leq \sum_{i=1}^N \lambda_i J(u^i)$  where  $\sum_{i=1}^N \lambda_i = 1$ . Proof by induction. For  $N = 1$  and since  $\lambda_1 = 1$ , then we have

$$J(u^1) = J(u^1)$$

The case for  $N = 2$  comes for free, from the definition of  $J$  being a convex function

$$J((1 - \lambda)u^1 + \lambda u^2) \leq (1 - \lambda)J(u^1) + \lambda J(u^2) \quad (\text{A})$$

By making  $(1 - \lambda) \equiv \lambda_1, \lambda \equiv \lambda_2$ , the above can be written as

$$J(\lambda_1 u^1 + \lambda_2 u^2) \leq \lambda_1 J(u^1) + \lambda_2 J(u^2)$$

We now assume it is true for  $N = k - 1$ . In other words, the inductive hypothesis below is given as true

$$J\left(\sum_{i=1}^{k-1} \lambda_i u^i\right) \leq \sum_{i=1}^{k-1} \lambda_i J(u^i) \quad (*)$$

Now we have to show it will also be true for  $N = k$ , which is

$$\begin{aligned} \sum_{i=1}^k \lambda_i J(u^i) &= \lambda_1 J(u^1) + \lambda_1 J(u^1) + \dots + \lambda_k J(u^k) \\ &= (1 - \lambda_k) \left( \frac{\lambda_1}{(1 - \lambda_k)} J(u^1) + \frac{\lambda_1}{(1 - \lambda_k)} J(u^1) + \dots + \frac{\lambda_{k-1}}{(1 - \lambda_k)} J(u^{k-1}) + \frac{\lambda_k}{(1 - \lambda_k)} J(u^k) \right) \\ &= (1 - \lambda_k) \left( \frac{\lambda_1}{(1 - \lambda_k)} J(u^1) + \frac{\lambda_1}{(1 - \lambda_k)} J(u^1) + \dots + \frac{\lambda_{k-1}}{(1 - \lambda_k)} J(u^{k-1}) \right) + \lambda_k J(u^k) \\ &= (1 - \lambda_k) \left( \sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} J(u^i) \right) + \lambda_k J(u^k) \end{aligned} \quad (1)$$

Now we take advantage of the inductive hypothesis Eq. (\*) on  $k - 1$ , which says that  $J\left(\sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} u^i\right) \leq \sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} J(u^i)$ . Using this in (1) changes it to  $\geq$  relation

$$\sum_{i=1}^k \lambda_i J(u^i) \geq (1 - \lambda_k) J\left(\sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} u^i\right) + \lambda_k J(u^k) \quad (2)$$

We now take advantage of the case of  $N = 2$  in (A) by viewing RHS of (2) as  $(1 - \lambda_k) J(u^1) + \lambda_k J(u^2)$ , where we let  $u^1 \equiv \sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} u^i, u^2 \equiv u^k$ . Hence we conclude that

$$(1 - \lambda_k) J\left(\sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} u^i\right) + \lambda_k J(u^k) \geq J\left((1 - \lambda_k) \sum_{i=1}^{k-1} \frac{\lambda_i}{(1 - \lambda_k)} u^i + \lambda_k u^k\right) \quad (3)$$

Using (3) in (2) gives (the  $\geq$  relation remains valid, even more now, since we replaced



something in RHS of (2), by something smaller)

$$\begin{aligned} \sum_{i=1}^k \lambda_i J(u^i) &\geq J\left((1-\lambda_k) \sum_{i=1}^{k-1} \frac{\lambda_i}{(1-\lambda_k)} u^i + \lambda_k u^k\right) \\ &= J\left(\left(\sum_{i=1}^{k-1} \lambda_i u^i\right) + \lambda_k u^k\right) \end{aligned}$$

Hence

$$\sum_{i=1}^k \lambda_i J(u^i) \geq J\left(\sum_{i=1}^k \lambda_i u^i\right)$$

QED.

## 0.5 Problem 5

### PROBLEM DESCRIPTION

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### ECE 719 – Homework Hessian

For  $u \in \mathbf{R}^n$ , define

$$J(u) = -(u_1 u_2 u_3 \cdots u_n)^{1/n}.$$

Prove that  $J(u)$  is convex on the positive orthant; i.e., the set defined by  $u_i > 0$  for  $i = 1, 2, \dots, n$ .

### SOLUTION

Assuming  $J(u)$  is twice continuously differentiable ( $C^2$ ) in  $u_1, u_2, \dots, u_n$ , then if we can show that the Hessian  $\nabla^2 J(u)$  is positive semi-definite on  $u_i > 0$ , then this implies  $J(u)$  is convex. The first step is to determine  $\nabla^2 J(u)$ .

$$\begin{aligned} \frac{\partial J}{\partial u_i} &= -\frac{1}{n} (u_1 u_2 \cdots u_n)^{\frac{1}{n}-1} \prod_{k=1, k \neq i}^n u_k = \frac{1}{n} \frac{J(u)}{(u_1 u_2 \cdots u_n)^{\frac{1}{n}}} \prod_{k=1, k \neq i}^n u_k = \frac{1}{n} \frac{J(u)}{\prod_{k=1}^n u_k} \prod_{k=1, k \neq i}^n u_k \\ &= \frac{1}{n} \frac{J(u)}{u_i} \end{aligned}$$

And

$$\begin{aligned} \frac{\partial^2 J}{\partial u_i^2} &= \frac{1}{n} \frac{\left(\frac{1}{n} \frac{J(u)}{u_i}\right)}{u_i} - \frac{1}{n} \frac{J(u)}{u_i^2} \\ &= \frac{1}{n^2} \frac{J(u)}{u_i^2} - \frac{1}{n} \frac{J(u)}{u_i^2} \\ &= \frac{1}{n} \frac{J(u)}{u_i^2} \left(\frac{1}{n} - 1\right) \end{aligned}$$

And the cross derivatives are

$$\begin{aligned} \frac{\partial^2 J}{\partial u_i \partial u_j} &= \frac{\partial}{\partial u_j} \left( \frac{1}{n} \frac{J(u)}{u_i} \right) \\ &= \frac{1}{n} \frac{J(u)}{u_i} \frac{1}{u_j} \\ &= \frac{1}{n^2} \frac{J(u)}{u_i u_j} \end{aligned}$$

Therefore

$$\nabla^2 J(u) = \begin{pmatrix} \frac{1}{n^2} \frac{J(u)}{u_1^2} (1-n) & \frac{1}{n^2} \frac{J(u)}{u_1 u_2} & \cdots & \frac{1}{n^2} \frac{J(u)}{u_1 u_n} \\ \frac{1}{n^2} \frac{J(u)}{u_2 u_1} & \frac{1}{n^2} \frac{J(u)}{u_2^2} (1-n) & \cdots & \frac{1}{n^2} \frac{J(u)}{u_2 u_n} \\ \vdots & \cdots & \ddots & \vdots \\ \frac{1}{n^2} \frac{J(u)}{u_n u_1} & \frac{1}{n^2} \frac{J(u)}{u_n u_2} & \cdots & \frac{1}{n^2} \frac{J(u)}{u_n^2} (1-n) \end{pmatrix}$$

Now we need to show that  $\nabla^2 J(u)$  is positive semi-definite. For  $n = 1$ , the above reduces to

$$\nabla^2 J(u) = \frac{J(u)}{u_1^2} (1-1) = 0$$

Hence non-negative. This is the same as saying the second derivative is zero. For  $n = 2$

$$\nabla^2 J(u) = \begin{pmatrix} \frac{1}{4} \frac{J(u)}{u_1^2} & \frac{1}{u_1 u_2} \frac{1}{4} J(u) \\ \frac{1}{u_2 u_1} \frac{1}{4} J(u) & \frac{1}{4} \frac{J(u)}{u_2^2} \end{pmatrix} = \begin{pmatrix} \frac{-1}{u_1^2} \frac{1}{4} J(u) & \frac{1}{u_1 u_2} \frac{1}{4} J(u) \\ \frac{1}{u_2 u_1} \frac{1}{4} J(u) & \frac{-1}{u_2^2} \frac{1}{4} J(u) \end{pmatrix}$$

The first leading minor is  $\frac{-1}{4u_1^2} J(u)$ , which is positive, since  $J(u) < 0$  and  $u_i > 0$  (given). The

second leading minor is

$$\Delta_2 = \begin{vmatrix} \frac{-1}{u_1^2} \frac{1}{4} J(u) & \frac{1}{u_1 u_2} \frac{1}{4} J(u) \\ \frac{1}{u_2 u_1} \frac{1}{4} J(u) & \frac{-1}{u_2^2} \frac{1}{4} J(u) \end{vmatrix} = 0$$

Hence all the leading minors are non-negative. Which means  $\nabla^2 J(u)$  is semi-definite. We will look at  $n = 3$

$$\nabla^2 J(u) = \begin{pmatrix} \frac{-2}{u_1^2} \frac{1}{9} J(u) & \frac{1}{u_1 u_2} \frac{1}{9} J(u) & \frac{1}{u_1 u_3} \frac{1}{9} J(u) \\ \frac{1}{u_2 u_1} \frac{1}{9} J(u) & \frac{-2}{u_2^2} \frac{1}{9} J(u) & \frac{1}{u_2 u_3} \frac{1}{9} J(u) \\ \frac{1}{u_3 u_1} \frac{1}{9} J(u) & \frac{1}{u_3 u_2} \frac{1}{9} J(u) & \frac{-2}{u_3^2} \frac{1}{9} J(u) \end{pmatrix}$$

The first leading minor is  $\frac{-2}{9u_1^2} J(u)$ , which is positive again, since  $J(u) < 0$  for  $u_i > 0$  (given).

And the second leading minor is  $\frac{1}{27} J^2 \frac{u^2}{u_1^2 u_2^2}$

which is positive, since all terms are positive. The third leading minor is

$$\Delta_3 = \begin{vmatrix} \frac{-2}{u_1^2} \frac{1}{9} J(u) & \frac{1}{u_1 u_2} \frac{1}{9} J(u) & \frac{1}{u_1 u_3} \frac{1}{9} J(u) \\ \frac{1}{u_2 u_1} \frac{1}{9} J(u) & \frac{-2}{u_2^2} \frac{1}{9} J(u) & \frac{1}{u_2 u_3} \frac{1}{9} J(u) \\ \frac{1}{u_3 u_1} \frac{1}{9} J(u) & \frac{1}{u_3 u_2} \frac{1}{9} J(u) & \frac{-2}{u_3^2} \frac{1}{9} J(u) \end{vmatrix} = 0$$

Hence non-of the leading minors are negative. Therefore  $\nabla^2 J(u)$  is semi-definite. The same pattern repeats for higher values of  $n$ . All leading minors are positive, except the last leading minor will be zero.

### 0.5.1 Appendix

Another way to show that  $\nabla^2 J(u)$  is positive semi-definite is to show that  $x^T (\nabla^2 J(u)) x \geq 0$  for any vector  $x$ . (since  $\nabla^2 J(u)$  is symmetric).

$$x^T (\nabla^2 J(u)) x = \begin{pmatrix} x_1 & x_2 & \cdots & x_n \end{pmatrix} \begin{pmatrix} \frac{1}{n} \frac{J(u)}{u_1^2} \left(\frac{1}{n} - 1\right) & \frac{1}{n^2} \frac{J(u)}{u_1 u_2} & \cdots & \frac{1}{n^2} \frac{J(u)}{u_1 u_n} \\ \frac{1}{n^2} \frac{J(u)}{u_2 u_1} & \frac{1}{n} \frac{J(u)}{u_2^2} \left(\frac{1}{n} - 1\right) & \cdots & \frac{1}{n^2} \frac{J(u)}{u_2 u_n} \\ \vdots & \cdots & \ddots & \vdots \\ \frac{1}{n^2} \frac{J(u)}{u_n u_1} & \frac{1}{n^2} \frac{J(u)}{u_n u_2} & \cdots & \frac{1}{n} \frac{J(u)}{u_n^2} \left(\frac{1}{n} - 1\right) \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

Now the idea is to set  $n = 1, 2, 3, \dots$  and show that the resulting values  $\geq 0$  always. For  $n = 1$ , we obtain 0 as before. For  $n = 2$ , let

$$\Delta = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} \frac{1}{n} \frac{J(u)}{u_1^2} \left(\frac{1}{n} - 1\right) & \frac{1}{n^2} \frac{J(u)}{u_1 u_2} \\ \frac{1}{n^2} \frac{J(u)}{u_2 u_1} & \frac{1}{n} \frac{J(u)}{u_2^2} \left(\frac{1}{n} - 1\right) \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}. \text{ Expanding gives}$$

$$\begin{aligned} \Delta &= \left( x_1 \frac{1}{n} \frac{J(u)}{u_1^2} \left(\frac{1}{n} - 1\right) + x_2 \frac{1}{n^2} \frac{J(u)}{u_2 u_1} \right) x_1 + \left( x_1 \frac{1}{n^2} \frac{J(u)}{u_1 u_2} + x_2 \frac{1}{n} \frac{J(u)}{u_2^2} \left(\frac{1}{n} - 1\right) \right) x_2 \\ &= x_1^2 \frac{1}{n} \frac{J(u)}{u_1^2} \left(\frac{1}{n} - 1\right) + x_1 x_2 \frac{1}{n^2} \frac{J(u)}{u_2 u_1} + x_2 x_1 \frac{1}{n^2} \frac{J(u)}{u_1 u_2} + x_2^2 \frac{1}{n} \frac{J(u)}{u_2^2} \left(\frac{1}{n} - 1\right) \\ &= x_1^2 \frac{1}{2} \frac{J(u)}{u_1^2} \left(\frac{1}{2} - 1\right) + x_1 x_2 \frac{1}{4} \frac{J(u)}{u_2 u_1} + x_2 x_1 \frac{1}{4} \frac{J(u)}{u_1 u_2} + x_2^2 \frac{1}{2} \frac{J(u)}{u_2^2} \left(\frac{1}{2} - 1\right) \end{aligned}$$

The RHS above becomes, and by replacing  $J(u) = -\sqrt{u_1 u_2}$  for  $n = 2$

$$\begin{aligned} -\frac{1}{4} x_1^2 \frac{J(u)}{u_1^2} + x_1 x_2 \frac{1}{2} \frac{J(u)}{u_2 u_1} - \frac{1}{4} x_2^2 \frac{J(u)}{u_2^2} &= \frac{1}{4} x_1^2 \frac{\sqrt{u_1 u_2}}{u_1^2} - x_1 x_2 \frac{1}{2} \frac{\sqrt{u_1 u_2}}{u_2 u_1} + \frac{1}{4} x_2^2 \frac{\sqrt{u_1 u_2}}{u_2^2} \\ &= \left( \frac{1}{\sqrt{4}} \frac{(u_1 u_2)^{\frac{1}{4}}}{u_1} x_1 - \frac{1}{\sqrt{4}} \frac{(u_1 u_2)^{\frac{1}{4}}}{u_2} x_2 \right)^2 \end{aligned}$$

Where we completed the square in the last step above. Hence  $x^T (\nabla^2 J(u)) x \geq 0$ . The same process can be continued for  $n$  higher. Hence  $\nabla^2 J(u)$  is positive semi-definite.