

1. Consider the nonlinear program:

$$\begin{aligned} \min \quad & 4(x_1 - 1)^2 + (x_2 - 1)^2 \\ \text{s.t.} \quad & 16x_1 + 6x_2 = 63 \end{aligned}$$

- (a) Form the Lagrangian function for this model.
- (b) Write the stationary conditions for the Lagrangian.
- (c) Solve your stationary condition for  $x_1$  and  $x_2$ . Explain why your answers are optimal for the original model.
- (d) Explain why the constraint  $x_1 \leq 2$  would be active if added to the original model.
- (e) Use the Lagrangian approach to compute an optimal solution to the model with the added constraint.

$$a) L(x; \lambda, \mu) = f_0(x) + \sum \lambda_i f_i(x) + \sum \mu_i h_i(x)$$

$$b) \left\{ \begin{array}{l} Pf) 16x_1 + 6x_2 - 63 = 0 \\ Df) \langle 8x_1 - 8 + 16\mu, 2x_2 - 2 + 6\mu \rangle = 0 \end{array} \right.$$

$$c) \begin{bmatrix} 8 & 0 & 16 \\ 0 & 2 & 6 \\ 16 & 6 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \mu \end{bmatrix} = \begin{bmatrix} 8 \\ 2 \\ 63 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ \mu \end{bmatrix} = \begin{bmatrix} 66/25 \\ 173/50 \\ -41/50 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 66/25 \\ 173/50 \end{bmatrix} \quad f_0 \text{ is convex as it is parabolas with positive coefficients}$$

$h(x)$  is affine

$$f(x^*) = 4(66/25 - 1)^2 + (173/50 - 1)^2 = \frac{1681}{100}$$

```
(%o11) m:matrix([8,0,16],[0,2,6],[16,6,0]);
v:matrix([8],[2],[63]);
m1:invert(m);
m1.v;
m
v
m1
(%o11)
```

$$\begin{aligned} m &= \begin{bmatrix} 8 & 0 & 16 \\ 0 & 2 & 6 \\ 16 & 6 & 0 \end{bmatrix} \\ v &= \begin{bmatrix} 8 \\ 2 \\ 63 \end{bmatrix} \\ m^{-1} &= \begin{bmatrix} \frac{9}{200} & -\frac{3}{25} & \frac{1}{25} \\ -\frac{3}{25} & \frac{6}{25} & \frac{3}{50} \\ \frac{1}{25} & \frac{3}{50} & -\frac{1}{50} \end{bmatrix} \\ m^{-1}v &= \begin{bmatrix} \frac{66}{25} \\ \frac{173}{50} \\ -\frac{41}{50} \end{bmatrix} \end{aligned}$$

Thus the problem is convex and the local min from the KKT solution is the global min

$$d) \frac{66}{25} > 2 \Rightarrow x_1^* = 2 \quad \text{so this would be an active constraint.}$$

$$e) Pf) x_1 - 2 \leq 0 \quad Df) \lambda \geq 0$$

$$\langle 8x_1 - 8 + \lambda + 16\mu, 2x_2 - 2 + 6\mu \rangle = 0$$

$$(5) \lambda(x_1 - 2) = 0$$

$$\text{Set } x_1 = 2 \quad 32 + 6x_2 = 63 \quad x_2 = \frac{31}{6}$$

$$\begin{aligned} 16 - 8 + \lambda + 16\mu &= 0 \\ \lambda &= -16\mu - 8 \end{aligned}$$

$$-16\mu - 8 \geq 0$$

$$\begin{aligned} \frac{31}{3} - 2 + 6\mu &= 0 \\ \frac{1}{3} - \frac{31}{18} &\approx -1.389 < -1/2 \end{aligned}$$

Thus  $(2, \frac{31}{6})$  is the new optimal solution as it fits the KKT conditions

$$f(x^*) = 4(2 - 1)^2 + (\frac{31}{6} - 1)^2 = \frac{769}{36}$$

2. Consider the nonlinear program:

$$\begin{array}{ll} \min & x_1 \\ \text{s.t.} & x_1^2 - x_2 = 0 \\ & 2x_1 - x_2 = 1 \end{array}$$

- (a) Form the Lagrangian function for this model.
- (b) Write the stationary conditions for the Lagrangian.
- (c) Solve your stationary condition for  $x_1$  and  $x_2$ .
- (d) Identify an optimal solution to this problem. Explain your reasoning.
- (e) Explain the relationship between your answers to the previous two parts.

$$a) L(x, \lambda, \mu) = x_1 + 0 + \mu_1(x_1^2 - x_2) + \mu_2(2x_1 - x_2 - 1)$$

$$b) \text{PF) } \begin{array}{l} x_1^2 - x_2 = 0 \\ 2x_1 - x_2 - 1 = 0 \end{array} \quad \text{DF) } \langle 1 + 2x_1\mu_1 + 2\mu_2, -\mu_1 - \mu_2 \rangle = 0$$

$$c) \underset{\text{CS)}{x_2 = x_1}$$

$$\begin{array}{l} x_1^2 - 2x_1 + 1 = 0 \\ (x_1 - 1)^2 = 0 \end{array} \quad x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

d) This is the only feasible solution so it is optimal

e) This has only one feasible result so by default it is optimal. The only one has many feasible points  
so we have to choose a number

3. For the following mathematical programs, state the KKT conditions and determine whether a KKT point is a global optimum

(a)

$$\begin{array}{ll} \max & 6x_1 + 40x_2 + 5x_3 \\ \text{s.t.} & x_1 \sin x_2 + 9x_3 \leq 2 \\ & e^{18x_1+3x_2} + 14x_3 \geq 50 \\ & x_1, x_2 \geq 0 \end{array}$$

(b)

$$\begin{array}{ll} \max & 7 \ln x_1 + 4 \ln x_2 + 11 \ln x_3 \\ \text{s.t.} & (x_1 + 2)^2 - x_1 x_2 + (x_2 - 7)^2 \leq 80 \\ & 5x_1 + 7x_3 = 22 \\ & x_1, x_2, x_3 \geq 1 \end{array}$$

$$g) \text{PF) } \begin{array}{l} x_1 \sin x_2 + 9x_3 - 2 \leq 0 \\ 50 - e^{18x_1+3x_2} - 14x_3 \leq 0 \\ -x_1 \leq 0 \\ -x_2 \leq 0 \end{array}$$

$$\text{DF) } \lambda \geq 0$$

$$\begin{array}{l} 6 - \lambda_1 \sin x_2 - 18 \lambda_2 e^{3x_2 + 18x_1} - \lambda_3 = 0 \\ 40 - \lambda_1 x_1 \cos x_2 - 3 \lambda_2 e^{3x_2 + 18x_1} - \lambda_4 = 0 \end{array}$$

$$5 - 9 \lambda_1 - 14 \lambda_2 = 0$$

$$c) \lambda_1(x_1 \sin x_2 + 9x_3 - 2) = 0$$

$$\lambda_2(50 - e^{18x_1+3x_2} - 14x_3) = 0$$

$$\lambda_3 x_1 = 0$$

$$\lambda_4 x_2 = 0$$

$x_1 \sin x_2 + 9x_3 - 2 \leq 0$  is not convex/concave so the KKT points are not necessarily global optimum

```
grad(var, e) := makelist(diff(e, var1), var1, var);
5) grad(var,e):=makelist( d -- , var1, var)
```

```
f0(x,y,z):=6*x+40*y+5*z;
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```
f1(x,y,z):=x*sin(y)+9*z-2;
```

```
f2(x,y,z):=50-e^(18*x+3*y)-14*z;
```

```
f3(x,y,z):=-x;
```

```
f4(x,y,z):=-y;
```

```
2) f0(x,y,z):=6*x+40*y+5*z
```

```
3) f1(x,y,z):=x*sin(y)+9*z-2
```

```
4) f2(x,y,z):=50 - %e^(18*x+3*y) + - 14 z
```

```
5) f3(x,y,z):=-x
```

```
6) f4(x,y,z):=-y
```

```
grad([x, y, z], f0(x,y,z));
```

```
grad([x, y, z], f1(x,y,z));
```

```
grad([x, y, z], f2(x,y,z));
```

```
grad([x, y, z], f3(x,y,z));
```

```
grad([x, y, z], f4(x,y,z));
```

```
7) [6,40,5]
```

```
8) [sin(y),x cos(y),9]
```

```
9) [ - (18 %e^(3 y + 18 x)), -(3 %e^(3 y + 18 x)), - 14 ]
```

```
10) [-1,0,0]
```

```
11) [0,-1,0]
```

$x_1 \sin x_2 + 9x_3 - 2 \leq 0$  is not convex/concave so  
The KKT points are not necessarily global optimum

b) Pf)  $(x_1+7)^2 - x_1 x_2 + (x_2-7)^2 - 80 \leq 0$

$$1 - x_1 \leq 0$$

$$1 - x_2 \leq 0$$

$$1 - x_3 \leq 0$$

$$5x_1 + 7x_2 - 22 = 0$$

Pf)  $\frac{3}{x_1} + 2\lambda_1(x_1+7) - \lambda_1 x_2 - \lambda_2 + 5\mu = 0$

$$\frac{4}{x_2} + 2\lambda_2(x_2-7) - \lambda_1 x_1 - \lambda_3 + 7\mu = 0$$

$$\frac{11}{x_3} - \lambda_3 + 7\mu = 0$$

$$\lambda_1 \geq 0$$

(S)  $\lambda_1((x_1+7)^2 - x_1 x_2 + (x_2-7)^2 - 80) = 0$

$$\begin{aligned}\lambda_2(1-x_1) &= 0 \\ \lambda_3(1-x_2) &= 0\end{aligned}$$

grad(var, e) := makelist(diff(e, var1), var1, var);

) grad(var, e) := makelist( $\frac{d}{d \text{var1}} e, \text{var1}, \text{var}$ )

f0(x,y,z) := 7\*log(x) + 4\*log(y) + 11\*log(z);

f1(x,y,z) := (x+2)^2 - x\*y + (y-7)^2 - 80;

f2(x,y,z) := 1 - x;

f3(x,y,z) := 1 - y;

f4(x,y,z) := 1 - z;

h(x,y,z) := 5\*x + 7\*z - 22;

) f0(x,y,z) := 7 log(x) + 4 log(y) + 11 log(z)

) f1(x,y,z) := (x+2)^2 - x\*y + (y-7)^2 - 80

) f2(x,y,z) := 1 - x

) f3(x,y,z) := 1 - y

) f4(x,y,z) := 1 - z

) h(x,y,z) := 5\*x + 7\*z - 22

grad([x, y, z], f0(x,y,z));

grad([x, y, z], f1(x,y,z));

grad([x, y, z], f2(x,y,z));

grad([x, y, z], f3(x,y,z));

grad([x, y, z], f4(x,y,z));

grad([x, y, z], h(x,y,z));

)  $\left[ \frac{7}{x}, \frac{4}{y}, \frac{11}{z} \right]$

) [2\*(x+2) - y, 2\*(y-7) - x, 0]

) [-1, 0, 0]

) [0, -1, 0]

) [0, 0, -1]

) [5, 0, 7]

The objective function is concave

$h$  is affine and  $x_i \geq 1$  is convex  
so we just need to check if  $f_i$  is convex

$$\nabla^2 f_i = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \quad \text{tr}(\nabla^2 f_i) = 4 \quad \det(\nabla^2 f_i) = 4 - 1 = 3$$

$\therefore \sigma_p(\nabla^2 f_i) = \{3, 1\} \Rightarrow \nabla^2 f_i$  is p.d.  $\Rightarrow f_i$  is convex

Thus the KKT points would give global optimum

4. We will show in this problem that redundant constraints in constraint optimization can be useful. Consider the non-convex optimization problem

$$\begin{aligned}\min_{x_1, x_2 \in \mathbb{R}} \quad & x_1 x_2 \\ \text{s.t. } & x_1^2 + x_2^2 \leq 1, x_1 \geq 0, \text{ and } x_2 \geq 0.\end{aligned}\tag{1}$$

and

$$\begin{aligned}\min_{x_1, x_2 \in \mathbb{R}} \quad & x_1 x_2 \\ \text{s.t. } & x_1^2 + x_2^2 \leq 1, x_1 x_2 \geq 0, x_1 \geq 0, \text{ and } x_2 \geq 0.\end{aligned}\tag{2}$$

The only difference between (1) and (2) is that there is a redundant constraint  $x_1 x_2 \geq 0$  in (2). Indeed, both problems have exactly the same feasible set and objective function, with optimal primal objective value 0.

- (a) List the KKT conditions of both problems.  
(b) Show that the optimal value of the dual problem to (1) is  $-1/2$ .  
(c) Show that the optimal value of the dual problem to (2) is 0.

9)

(1)

Pf)  $x_1^2 + x_2^2 - 1 \leq 0$   
 $-x_1 \leq 0$   
 $-x_2 \leq 0$

(2)

$$x_1^2 + x_2^2 - 1 \leq 0$$
  
 $-x_1 \leq 0$   
 $-x_2 \leq 0$

grad(var, e) := makelist(diff(e, var1), var1, var);

) grad(var, e) := makelist( $\frac{d}{d \text{var1}} e, \text{var1}, \text{var}$ )

f0(x,y,z) := x\*y;

f1(x,y,z) := x^2 + y^2 - 1;

$$Pf) \quad x_1^2 + x_2^2 - 1 \leq 0$$

$$\begin{aligned} -x_1 &\leq 0 \\ -x_2 &\leq 0 \end{aligned}$$

$$x_1^2 + x_2^2 - 1 \leq 0$$

$$\begin{aligned} -x_1 &\leq 0 \\ -x_2 &\leq 0 \\ -x_1 x_2 &\leq 0 \end{aligned}$$

$$Df) \quad \lambda \geq 0$$

$$x_2 + 2\lambda_1 x_1 - \lambda_2 = 0$$

$$x_1 + 2\lambda_2 x_2 - \lambda_3 = 0$$

$$CS) \quad \lambda_1(x_1^2 + x_2^2 - 1) = 0$$

$$-\lambda_2 x_1 = 0$$

$$-\lambda_3 x_2 = 0$$

$$\lambda \geq 0$$

$$x_2 + 2\lambda_1 x_1 - \lambda_2 - \lambda_4 x_2 = 0$$

$$x_1 + 2\lambda_2 x_2 - \lambda_3 - \lambda_4 x_1 = 0$$

$$\lambda_1(x_1^2 + x_2^2 - 1) = 0$$

$$-\lambda_2 x_1 = 0$$

$$-\lambda_3 x_2 = 0$$

$$-\lambda_4 x_1 x_2 = 0$$

grad([x, y, z], f0(x,y,z)); grad([x, y, z], f1(x,y,z));

f0(x,y,z):=x\*y;

f1(x,y,z):=x^2+y^2-1;

f2(x,y,z):=-x;

f3(x,y,z):=-y;

f4(x,y,z):=-x\*y;

) f0(x,y,z):=x\*y

) f1(x,y,z):=x^2+y^2-1

) f2(x,y,z):=-x

) f3(x,y,z):=-y

) f4(x,y,z):=-x\*y

grad([x, y, z], f0(x,y,z));

grad([x, y, z], f1(x,y,z));

grad([x, y, z], f2(x,y,z));

grad([x, y, z], f3(x,y,z));

grad([x, y, z], f4(x,y,z));

) [y,x,0]

) [2x,2y,0]

) [-1,0,0]

) [0,-1,0]

) [-y,-x,0]

$$b) \quad g(\lambda) = \inf L(x; \lambda) = \inf (x_1 x_2 + \lambda_1 (x_1^2 + x_2^2 - 1) - \lambda_2 x_1 - \lambda_3 x_2) \\ = -\lambda_1$$

$$x_2 = \lambda_2 - 2\lambda_1 x_1$$

$$x_1 + 2\lambda_1 \lambda_2 - 4\lambda_1^2 x_1 - \lambda_3 = 0$$

$$x_1 = \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2}$$

$$x_2 = \lambda_2 - 2\lambda_1 \left( \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2} \right) = \frac{\lambda_2(1 - 4\lambda_1^2) - 2\lambda_1 \lambda_2 + 4\lambda_1^2 \lambda_2}{1 - 4\lambda_1^2} = \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2}$$

$$\left( \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2} \right)^2 + \left( \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2} \right)^2 \leq 1$$

$$(\lambda_2 - 2\lambda_1 \lambda_2)^2 + (\lambda_2 - 2\lambda_1 \lambda_2)^2 \leq (1 - 4\lambda_1^2)^2$$

$$\min (\lambda_2 - 2\lambda_1 \lambda_2)^2 + (\lambda_2 - 2\lambda_1 \lambda_2)^2 \leq (1 - 4\lambda_1^2)^2$$

$$0 \leq 1 - 4\lambda_1^2$$

$$\Rightarrow \lambda_1 \leq \frac{1}{2}$$

$$\therefore \max g(\lambda) = -\frac{1}{2}$$

$$g(\lambda) = \inf L(x; \lambda) = \inf (x_1 x_2 + \lambda_1 (x_1^2 + x_2^2 - 1) - \lambda_2 x_1 - \lambda_3 x_2 - \lambda_4 x_1 x_2) \\ = -\lambda_1$$

$$\max g(\lambda) = \max -\lambda_1 = -\min \lambda_1$$

$$\left( \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2} \right) \left( \frac{\lambda_2 - 2\lambda_1 \lambda_2}{1 - 4\lambda_1^2} \right) \geq 0$$

$$\lambda_2 \lambda_2 - 2\lambda_1 \lambda_2^2 - 2\lambda_1 \lambda_2^2 - 4\lambda_1^2 \lambda_2 \lambda_2 \geq 0$$

$$4\lambda_1^2\lambda_2\lambda_3 + 2\lambda_1(\lambda_2^2 + \lambda_3^2) - \lambda_2\lambda_3 < 0$$

$$-2(\lambda_2^2 + \lambda_3^2) \pm \sqrt{4(\lambda_2^2 + \lambda_3^2)^2 + 16\lambda_2^2\lambda_3^2} \geq \lambda_1 \\ 8\lambda_2\lambda_3$$

$$\min -2(\lambda_2^2 + \lambda_3^2) \pm \sqrt{4(\lambda_2^2 + \lambda_3^2)^2 + 16\lambda_2^2\lambda_3^2} \geq \min \lambda_1 \\ 8\lambda_2\lambda_3$$

$$0 \geq \min \lambda_1 \geq 0$$

$$\Rightarrow \min \lambda_1 = 0$$

$$\max g(\lambda) = -\min \lambda_1 = -0 = 0$$