Part II: Lagrange Multiplier Method & Karush-Kuhn-Tucker (KKT) Conditions

KKT Conditions

General Non-Linear Constrained Minimum:

Min: f[x]

Constrained by: h[x] = 0 (m equality constraints)

 $g[x] \le 0$ (k inequality constraints)

Introduce slack variables s_i for the inequality contraints: $g_i[x] + s_i^2 == 0$ and construct the monster Lagrangian:

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L[x,\lambda,\mu] = f[x] + \lambda h[x] + \sum \mu_i (g_i[x] + s_i^2)
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Recall the geometry of the Lagrange multiplier conditions: The gradient of the objective function must be orthogonal to the tangent plane of the (active) constraints. That is the projection of the gradient of f onto the space of directions tangent to the constraint "surface" is zero. The KKT conditions are analogous conditions in the case of constraints.

The KKT conditions are the following:

- 1) Gradient of the Lagrangian = 0
- 2) Constraints: h[x] = 0 (m equality constraints) & $g[x] \le 0$ (k inequality constraints)
- 3) Complementary Slackness (for the s_i variables) μ .s == 0
- 4) Feasibility for the inequality constraints: $s_i^2 \ge 0$
- 5) Sign condition on the inequality multipliers: $\mu \ge 0$

One final requirement for KKT to work is that the gradient of f at a feasible point must be a linear combination of the gradients for the equality constraints and the gradients of the active constraints: this is often called regularity of a feasible point.

At a feasible point for the constraints, the active constraints are those components of g with $g_i[x] = 0$ (if the value of the constraining function is < 0, that constraint is said to be inactive).

The solution of a set of KKT equations proceeds by cases according to which inequality constraints are Active & Inactive.

Example: Chong Zak Example 20.2

Consider a circuit with a 20V battery and two resistances in series: R and 10 ohm. We will investigate the maximization of the power in each of the resistors separately

■ Subproblem One: Max in resistor R

Maximize power absorbed by resistance R

Equivalently: minimized: - Power absorbed by $R = -i^2 R = -400 R/(10 + R)^2$ subject to $-R \le 0$.

The lagrangian is $L[R,\lambda] =$

$$L[R_{, \mu_{]}} := \frac{-400 R}{(10 + R)^{2}} + \mu (-R)$$

$$D[L[R, \mu], R]$$

$$D[L[R, \mu], \mu]$$

$$\frac{800 R}{(10 + R)^{3}} - \frac{400}{(10 + R)^{2}} - \mu$$

KKT also gives us the complementary slackness: μ .R = 0 and the sign condition for the inequality constraints: $\mu \ge 0$. But, if $\mu > 0$, then R==0 which is gives no power absorbed by R.

If the constraint is inactive we must solve

Solve[(D[L[R,
$$\mu$$
], R] /. { $\mu \rightarrow 0$ }) = 0, R] {{R \rightarrow 10}}

This gives the maximum

-R

■ Subproblem Two: Maximization of power in 10 ohm resistor

Find the value of R such that maximal power is delivered to the 10 ohm resistor, i.e:

Min: $- \frac{4000}{(10+R)^2}$ subject to $-R \le 0$

$$L[R_{-}, \mu_{-}] := \frac{-4000}{(10 + R)^2} + \mu (-R)$$

Again, KKT gives us a complementary slackness condition: μ .R = 0 and the sign condition for the inequality constraints: $\mu \ge 0$.

But, if $\mu = 0$, we must solve

Solve[(D[L[R,
$$\mu$$
], R]/. { $\mu \rightarrow 0$ }) == 0, R]

This first case has no solutions. However, R = 0 is possible in the case $\mu > 0$

KKT conditions for a Linear Program: (Franklin: p. 197)

■ Lagrangian

Min: c^T .x subject to the linear constraints:

$$A.x == b, x \ge 0,$$

the associated Lagrangian is:

$$L[x,\lambda] = c^T \cdot x + \text{Transpose}[\lambda] (A.x - b) + \mu \cdot (-x + s^{[2]})$$

The KKT conditions yield:

$$c^T$$
 - Transpose[λ].A + Transpose[μ] == 0
A.x - b ==0
x \ge 0,
2 Transpose[μ].s == 0
 μ \ge 0

Eliminating μ , we obtain the following: c^T - Transpose[λ]. A ≤ 0 , $(c^T$ - Transpose[λ]. A).x= - Transpose[μ]). $(s^{[2]}) = 0$ yielding

 c^T .x = Transpose[λ].A.x= Transpose[λ].b which is the **equilibrium condition** in mild disquise!

Example: Pedregal Example 3.10, p. 82

A certain electrical networks is designed to supply power x_i thru 3 channels.

$$p[x_{,} y_{,} z_{]} = z + (x^2 + y^2 + z^2/10)/2;$$

Contraint: $h[x,y,z]=x+y+z==5, x,y,z \ge 0$

Note that this gives a "slanted" triangles with vertices on the axes 5 units from the origin.

The KKT conditions give:

- 1) $\nabla f + \lambda \nabla h + \mu \nabla g = \{x,y,1+z/10\} + \lambda \{1,1,1\} + \{\mu_1,\mu_2,\mu_3\} == \{0,0,0\}$
- 2) Constraint: h==5
- 3) $\mu_1 x = 0$, $\mu_2 y = 0$, $\mu_3 z = 0$,

Checking for active constraints will divide in the consideration of 8 cases for the three inequality constraints.

KKT: Necessary Conditions for Quad. Program

■ Review: Quadratic Programs

The general quadratic program proposes to minimize an objective function of the form: Min: x.Q.x/2 + p.x subject to the linear constraints:

$$A.x == b, x \ge 0$$

Note that we may assume that Q is a symmetric matrix (and that Q is the Hessian of the objective function.) This is a non-linear program problem, for the objective function is a quadratic function (if Q is non-zero.)

■ Wolfe's Reduction to LP

Given the Quadratic Program as above, the associated Linear Program is:

$$A.x == b$$

$$Q.x + Transpose[A].u - v == -p$$

$$x \ge 0, v \ge 0, x.v == 0$$

Note that the vector u is not necessarily positive. The only additional feature is the **exclusion rule**: x.v == 0, this requirement states that the i^{th} components of x and v can not both be positive simultaneously.

Finally, note that the existence of a solution of the associated LP is only necessary for the existence of a solution to the QP and is sufficient in case Q is positive semi-definite.

In practice we solve the LP program:

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A.x==b

Q.x + Transpose[A].u -v + D.z == -p

x \ge 0l, u free, v \ge 0, z \ge 0

Min: \{1, ..., 1\}.z
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Initial Feasible Solution: x, u=0, v=0, z where x is a basic feasible solution of A.x ==b, $x \ge 0$, D is a diagonal matrix with entries ± 1 to correct the signs of z and z is a chosen such that Q.x + D.z == -p, $z \ge 0$.

KKT conditions for a Quadratic Program

■ Lagrangian

Given: x.Q.x/2 + p.x subject to the linear constraints: $A.x == b, x \ge 0,$

the associated Lagrangian is:

$$L[x,\lambda] = x.Q.x/2 + p.x + \lambda (A.x - b) + \mu.x$$

The KKT conditions yield:

Q.x + p - Transpose[A].
$$\lambda + \mu == 0$$

Transpose[λ]. $(A.x - b) == 0$
Transpose[μ]. $x == 0$
A. $x == b, x \ge 0$,

Exercises

I. Write down the KKT conditions for the problem:

Min f[x] =
$$-x_1^3 + x_2^2 - 2x_1x_3^2$$
 subject to the constraints:
 $2x_1 + x_2^2 + x_3 - 5 == 0$
 $5x_1^2 - x_2^2 - x_3 \ge 2$
 $x_i \ge 0$ for $i = 1,2,3$.

Verify that the KKT conditions are satisfied at (1,0,3).

II. Write down the KKT conditions for the problem:

Min f[x] =
$$x_1^2 + x_2^2 + x_3^2$$
 subject to the constraints:
- $x_1 + x_2 - x_3 \ge -10$
 $x_1 + x_2 + 4$ $x_3 \ge 20$

Find all the solutions.

III. Here's an exercise from Grieg p. 142):

Minimize:
$$f[x] = (x_1 - 1)^2 + (x_2 - 2)^2 + (x_3 - 3)^2 + (x_4 - 4)^2$$

subject to:
$$-x_1 - x_2 - x_3 - x_4 + 5 \ge 0$$
$$-3x_1 - 3x_2 - 2x_3 - x_4 + 10 \ge 0$$
$$x \ge 0$$

Verify the KT conditions and then expand f to obtains a Quadratic Program and compare the KT conditions with Wolfe's.

Solve the problem using Wolfe's method.

Bonus: Solve the problem using a projected gradient methods (mimic affine scaling) and a projected Newton's method. Compare the results.