## Practice Problems 14 - Solutions: Constrained optimization and Convex sets

## **EXERCISES**

1. \*Let  $f: \mathbb{R}^2 \to \mathbb{R}$  be defined as

$$f(x,y) = -(x - \alpha)^2 - (y - \alpha)^2$$

Consider the following optimization problem parametrized by  $\alpha \in \mathbb{R}$ 

$$\max_{x,y} f(x,y)$$

subject to the constraint

$$\mathcal{D} = \{(x, y) \in \mathbb{R}^2 : xy \le 1\}$$

(a) Explain why this optimization problem has a solution (an intuitive explanation suffices). Is a solution guaranteed if instead it was a minimization problem?

Answer: The objective function is continuous, and the feasible set is closed. Furthermore, we know that the solutions must live in a bounded subset of the feasible set because the function decreases when x and y grow further apart from  $\alpha$ . A formal way to say this is that (x,y)=(0,0) is feasible and any point with a larger distance with respect to  $(\alpha,\alpha)$  should be less desirable, this is, the solution must satisfy that  $\|(x,y)\| \leq \|(\alpha,\alpha)\|$ . By taking the intersection of this set and the feasible set we have that it is bounded and closed, thus compact, and Weierstrass ensures the existence of a solution. This will no longer be true if it is a minimization because it is always feasible to decrease the objective function by taking a point with a larger distance from  $(\alpha,\alpha)$ .

(b) Is the Qualification Constraint of the Theorem of Kuhn-Tucker satisfied?

**Answer:** Yes, if we happen to be in a situation where the constrain is binding, then  $x, y \neq 0$  and the jacobian is  $DG(\cdot) = [y \ x]$  which has rank 1 as desired.

(c) Write the Lagrangean and the Kuhn-Tucker conditions. Denote the multiplier by  $\lambda$ .

$$\mathcal{L}(x, y, \lambda) = -(x - \alpha)^2 - (y - \alpha)^2 + \lambda(1 - xy)$$

$$x] -2(x - \alpha) = \lambda y \tag{1}$$

$$y] -2(y - \alpha) = \lambda x \tag{2}$$

$$\lambda] xy \le 1 (3)$$

$$cs] \quad \lambda(1 - xy) = 0 \tag{4}$$

(d) Argue that the analysis can be split in three cases:  $\lambda = 0, 2$  and all other lambdas. **Answer** From 4 we see that there is a case when  $\lambda = 0$ , and from 1 and 2 we see that if  $\lambda = 2$  those two conditions are the same, so that is another case and when all of the other possible  $\lambda$ 's are the third case.

(e) In each case impose conditions on  $\alpha$  to ensure the existence of  $(x,y) \in \mathbb{R}^2$  that satisfies the Kuhn-Tucker conditions, and the value (if any) for which the constraint is active.

**Answer:** Case 1:  $\lambda = 0$ , then from 1 and 2  $x = y = \alpha$  and from 3  $\alpha^2 \le 1$  i.e. we need  $|\alpha| \le 1$ .

Case 2:  $\lambda = 2$ . From 4 we know that 3 binds and by combining it with 1 we have that x = 1/y and  $x^2 - \alpha x + 1 = 0$ , so  $x = \frac{\alpha \pm \sqrt{\alpha^2 - 4}}{2}$  so we need  $|\alpha| \ge 2$ .

Case 3:  $\lambda \notin \{0,2\}$ , By subtracting 2 from 1 we learn that  $(x-y)(2-\lambda)=0$ . So x=y and from 3,  $x=\pm 1$  and so  $\lambda=2(\alpha-1)$  when x=1 for which we need a>1 and  $\lambda=-2(\alpha+1)$  hence we need  $\alpha<-1$ . Then in general for this case to work we need  $|\alpha|>1$ .

(f) Assume that given some  $\alpha$ , there exists a global max  $(x^*, y^*)$  where the constraint is effective and with associated multiplier  $\lambda^*$ . What is the interpretation of  $\lambda^*$ . What do we know about the multiplier if the constrain is not active?

**Answer:**  $\lambda^*$  approximates the increase in the objective function when the constraint is relaxed. In this case  $xy \leq 1$  changing to  $xy \leq 1 + \epsilon$  for example. If the constraint is not active, relaxing it, should not change the local max found, and thus the value objective, Therefore,  $\lambda^* = 0$ .

(g) Describe the optimal solution of the maximization problem as a function of  $\alpha$ .

## Answer:

- If  $|\alpha| \leq 1$ , the global max is  $(\alpha, \alpha)$ .
- If  $\alpha < -1$  the global max is (-1, -1).
- If  $1 < \alpha$  the global max is (1,1)
- 2. Billy optimizes a  $C^1$  quasi-concave utility with respect to cheese curds and brats u(c, b). He can spend at most \$50 on these goods, and wants to buy at least 20 units combined in order to support the industry. Keep in mind that, of course, he cannot buy or eat negative quantities.
  - (a) What are the minimal conditions on the parameters or on the utility function to ensure the Kuhn-Tucker theorem applies for all critical points we might find.

**Answer:** We only need to have a non-empty set:  $20 \le \min\{50/P_c, 50/P_b\}$ . and if  $P_c = P_b = 5/2$ , we must re-write the problem as one with one equality constraints: b + c = 20 and non-negativity constraints.

(b) Assuming that the utility satisfies local non-satiation, that a solution exists, and that the price of cheese curds is smaller than the price of brats:  $P_c < P_b$  what are the possible combination of constraints that can bind?

**Answer:** Lets call the non-negativity on c the constraint c, similarly for b, the budget constraint, bc and the last constraint, s for support. With local non-satiation, we can rule out solutions in the interior of the feasible set, you can have binding the following

combinations:  $\{c\}, \{c, s\}, \{c, bc\}, \{c, bc, s\}, \{bc\}, \{bc, s\}, \{bc, b\}, \{bc, b, s\}, \{b\}, \{b, s\}$  and  $\{s\}.$ 

(c) Non-negativity constraints are usually dealt with a slightly different formulation: they are not added to the Lagrangean; instead the conditions are that  $\partial \mathcal{L}/\partial x \leq 0$  for any variable, x, with non-negativity constraints and a complementary slackness condition that  $x(\partial \mathcal{L}/\partial x) = 0$ . Show that the two formulations are equivalent.

**Answer:** Let the traditional Kuhn Tucker conditions be denoted by:  $\partial \mathcal{L}'/\partial x$ , we know they must be equal to zero. By arbitrarily eliminating the multiplier associated with the non-negativity constraint, we conclude the new condition on the lagrange must be that  $\partial \mathcal{L}/\partial x \leq 0$  as desired. This inequality holds with equality whenever the removed multiplier is zero, and this multiplier is zero if x > 0. Similarly, if the inequality is strict it must be because the restriction binds, and so x = 0. These two factors are summarized in the complementary slackness condition that states that the restriction binds, x = 0 or the modified condition binds,  $\partial \mathcal{L}/\partial x = 0$ .

## 3. Show that the following sets are convex

(a) \*The set of functions whose integral equals 1

**Answer:** Let  $f \neq g$  be two functions in the set, and let  $h = \lambda f + (1 - \lambda)g$  with  $\lambda \in (0,1)$  then  $\int h = \lambda \int f + (1 - \lambda) \int g = 1$ .

(b) \*The set of positive definite matrices

**Answer:** Take an arbitrary  $x \neq 0$  and two pd matrices, A, B. For  $C = \lambda A + (1 - \lambda)B$  we have that  $x'Cx = \lambda x'Ax + (1 - \lambda)x'Bx > 0$ .

(c) Any set of the form  $\{x \in X : G(x) \leq 0\}$  where  $G: X \to \mathbb{R}$  is affine.

**Answer:** Take arbitrary  $x_1, x_2$  in the set, then for  $x_{\lambda}$  being the linear combination of those two in the usual way we have  $G(x_{\lambda}) = ax_{\lambda} + b = \lambda ax_1 + (1 - \lambda)ax_2 + b \le G(x_1) + G(x_2)$  with equality if b = 0, and  $G(x_1) + G(x_2) \le 0$ .

(d) \*The cartesian product of 2 convex sets.

**Answer:** Denote the two sets by A and B, let  $a_1, a_2 \in A$  and  $b_1, b_2 \in B$ . then  $(a_{\lambda}, b_{\lambda})$  is the linear combination of  $(a_1, b_1) \in A \times B$  and  $a_2, b_2) \in A \times B$  which is also in the cartesian product because it is coordinate-by-coordinate.

(e) Any vector space

**Answer:** By definition for any vector space X  $\lambda x + (1 - \lambda)y \in X$  so long as  $\lambda$  is a scalar (and to define a convex combination we only need it to have a norm strictly less than 1) and x, y are in the set.

(f) The set of contraction mappings

**Answer:** Let h be the convex combination of two contractions, f, g. Then  $h(x) - h(y) = \lambda(f(x) - f(y)) + (1 - \lambda)(g(x) - g(y)) \le \lambda \beta_f(x - y) + (1 - \lambda)\beta_g(x - y) = (\lambda \beta_f + (1 - \lambda)\beta_g)(x - y) = \beta_h(x - y)$  where  $\beta_f, \beta_g < 1$ , so  $\beta_h < 1$ ; hence h is also a contraction.

(g) \*Supermodular functions If f and g are SPM:

$$f(x \lor x') + f(x \land x') \ge f(x) + f(x')$$

$$g(x \lor x') + g(x \land x') \ge g(x) + g(x')$$

By multiplying both sides of the first inequality by  $\lambda$ , both sides of the other by  $1 - \lambda$  and adding them together, we obtain the desired resilt.

4. Give an example of a set of functions that is not convex

**Answer:** Consider the set of discontinuous functions.

5. The set of invertible matrices is not convex, provide a counterexample to show this.

**Answer:** Let A be an invertible matrix, then -A is also invertible, but the linear combination with  $\lambda = 1/2$  is the zero matrix, which is not invertible.

6. Are finite intersections of open sets in  $\mathbb{R}^n$  convex?

**Answer:** No, consider  $\mathbb{R}$  open and  $(-1,0) \cup (1,2)$  open. Note that the second is equal to the intersection of these two sets and is not convex.

7. Show that the set of sequences in  $\mathbb{R}^n$  that posses a convergent subsequent is not a convex set

**Answer:** Take  $x_n = 1$  if n is odd and  $x_n = n$  if n is even and  $y_n = 1$  if n is even and  $y_n = n$  if n is odd.