

# Homework 4: Convex Functions

## Instructions

Please submit a .qmd file along with a rendered pdf to the Brightspace page for this assignment. You may use whatever language you like within your qmd file, I recommend python, julia, or R.

## Problem 1: (Exercise 3.2 in CVX Book)

- (a) The figure below shows three levels sets for a function  $f$ . The value of the function on the level set is indicated by the number next to each curve. For example, the curve labeled 1 corresponds to the points  $\mathbf{x} \in \mathbb{R}^2$  satisfying  $f(\mathbf{x}) = 1$ . Determine whether it is possible for the function  $f$  to be convex, concave, quasiconvex, or quasiconcave. Give a brief justification for your answer. Note: it may be that several options are possible, that one is possible, or that none at all are.

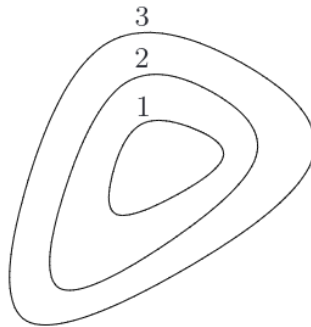


Figure 1: Level Curves of  $f$

- (b) The figure below shows level sets for a different function  $g$ . Again determine whether it is possible for the function  $g$  to be convex, concave, quasiconvex, or quasiconcave. Note: it may be that several options are possible, that one is possible, or that none at all are.



Figure 2: Level Curves of  $g$

### Problem 2: (CVX Book Extended Exercises 3.1)

Determine the curvature of the functions below. Your responses can be affine, convex, concave, or none (meaning not convex or concave). Provide a brief justification

- (a)  $f(x) = \min(2, x, \sqrt{x})$  with  $\text{dom} f = \{x \mid x \geq 0\}$  (i.e.  $\mathbb{R}_+$ )
- (b)  $f(x) = x^3$ , with  $\text{dom} f = \mathbb{R}$
- (c)  $f(x) = x^3$ , with  $\text{dom} f = \{x \mid x \geq 1\}$
- (d)  $f(x, y) = \sqrt{x \min(y, 2)}$ , with  $\text{dom} f = \{(x, y) \mid x \geq 0, y \geq 0\}$  (i.e.  $\mathbb{R}_+^2$ )
- (e)  $f(x, y) = (\sqrt{x} + \sqrt{y})^2$  with  $\text{dom} f = \{(x, y) \mid x \geq 0, y \geq 0\}$  (i.e.  $\mathbb{R}_+^2$ )

### Problem 3: (Selected from CVX Book Extended Exercises 3.51-3.52)

For each of the following problems implement the following functions using Disciplined Convex Programming and CVX, and use CVX to verify that they are convex

- (a)  $f(x, y) = \frac{1}{xy}$ , with  $\text{dom} f = \mathbb{R}_{++}^2$ . Hint: Use the Atoms listed below part (b) as well as addition, subtraction, and scalar multiplication. There are multiple ways to solve this problem.
- (b)  $f(x, y) = \sqrt{1 + \frac{x^4}{y}}$ , with  $\text{dom} f = \mathbb{R} \times \mathbb{R}_{++}$  (this means  $x$  is any real number and  $y$  is strictly greater than 0).

Hint: The following atoms may be helpful, there are multiple ways to solve this problem.

- `inv_pos(u)`, which is  $1/u$ , with domain  $\mathbb{R}_{++}$
- `square(u)`, which is  $u^2$ , with domain  $\mathbb{R}$
- `sqrt(u)`, which is  $\sqrt{u}$ , with domain  $\mathbb{R}_+$
- `geo_mean(u, v)`, which is  $\sqrt{uv}$ , with domain  $\mathbb{R}_+^2$

- `quad_over_lin(u,v)`, which is  $u^2/v$ , with domain  $\mathbb{R} \times \mathbb{R}_{++}$
- `norm2(u,v)`, which is  $\sqrt{u^2 + v^2}$ , with domain  $\mathbb{R}^2$ .

#### Problem 4: Periodic Poisson Regression to predict Car Crashes

For this problem, we will be working with the dataset [nyc\\_crashes.csv](#), which contains a time series of the number of car crashes occurring in Manhattan every hour for a period of time. Here we will develop a Poisson regression model to predict rate of crashes during each time of day.

- (a) Consider the following statistical model for the number of crashes  $N_i$  during hour  $i$  of a given day.

$$N_i \sim \exp(-\lambda_i) \frac{\lambda_i^{N_i}}{N_i!}$$

where  $i$  ranges from 0 to 23 and corresponds to the hour of the day. Suppose that we have a dataset of counts for crashes where  $C_{ni}$  is the number of crashes that occur in the  $i$ th hour of the  $n$ th day. Then the log likelihood function for the parameters  $\lambda$  is:

$$-\log(p(C|\lambda)) = \sum_{n=1}^N \sum_{i=0}^{23} (\lambda_i - C_{ni} \log(\lambda_i) + \log(C_{ni}!))$$

However, we can drop terms that don't depend on  $\lambda$  because they will have no impact on the maximum likelihood solution. This lets us form a simpler objective function:

$$L(C|\lambda) = \sum_{n=1}^N \sum_{i=0}^{23} (\lambda_i - C_{ni} \log(\lambda_i))$$

or if we use matrix-vector notation:

$$L(C|\lambda) = N\mathbf{1}^T \lambda - \mathbf{1}^T C \log(\lambda)$$

where  $\log(\lambda)$  is interpreted as a vector whose  $i$ th entry is  $\log(\lambda_i)$  and the vector  $\mathbf{1}^T$  has all entries equal to 1 and has the right dimension in each case to make the resulting expressions scalars (24 and  $N$ , for our dataset  $N$  will be 43).

Show that  $L(C|\lambda)$  is a convex function of the coefficients  $\lambda$  on the domain  $\lambda \in \mathbb{R}_{++}^{24}$ .

- (b) The log-likelihood function  $L$  can be minimized to find the maximum likelihood estimate for the  $\lambda$  coefficients. Formulate this constrained optimization problem in CVX and solve it for the NYC crashes dataset, that is solve:

$$\min_{\lambda} L(C|\lambda) \lambda \in \mathbb{R}_+^{24}$$

Make a plot of the  $\lambda$  coefficients

- (c) It would be reasonable to expect that the rate of crashes in two adjacent hours should be similar, i.e. the observation that there has been a crash at 3:01PM should influence the estimate of the rate of crashes between 2:00PM and 3:00PM in addition to between 3:00PM and 4:00PM. One way to implement this is through the following regularization term, which applies a penalty proportional to the square of the difference between  $\lambda_i$  and  $\lambda_{i+1}$ :

$$L_{pen}(C|\lambda, \rho) = L(C|\lambda) + \rho \left( \sum_{i=1}^{23} (\lambda_i - \lambda_{i+1})^2 + (\lambda_0 - \lambda_{23})^2 \right)$$

Assuming that  $\rho > 0$ , show that  $L_{pen}(C|\lambda)$  is a convex function on  $\mathbb{R}_{++}^{24}$

- (d) Formulate the regularized maximum likelihood problem:

$$\min_{\lambda} L_{pen}(C|\lambda, \rho) \lambda \in \mathbb{R}_+^{24}$$

and solve it using **CVX**. Solve it for a range of positive values  $\rho$  such that for your smallest values the solution appears like your solution to (b) and for your largest values the  $\lambda$  show much less variation over time.

Hint: There are many ways to implement this penalty term in **cvx**. A way that you might find useful is to define a matrix  $S$  where  $S_{ii} = 1$ ,  $S_{i,i-1} = -1$  for  $i > 1$  and  $S_{1,24} = -1$ , with  $S = 0$  for all other entries:

Then  $S\lambda = [(\lambda_1 - \lambda_{24}) \quad (\lambda_2 - \lambda_1) \quad \cdots \quad (\lambda_{24} - \lambda_{23})]^T$ , and you can use **cvx.square** and **cvx.sum** to construct the objective.