

2.15: Appendix II- Legendre Transformations

A *convex function* of a single variable $f(x)$ is one for which $f''(x) > 0$ everywhere. The *Legendre transform* of a convex function $f(x)$ is a function $g(p)$ defined as follows. Let p be a real number, and consider the line $y = px$, as shown in Figure 2.15.1. We define the point $x(p)$ as the value of x for which the difference $F(x, p) = px - f(x)$ is greatest. Then define $g(p) = F(x(p), p)$.²⁶ The value $x(p)$ is unique if $f(x)$ is convex, since $x(p)$ is determined by the equation

$$f'(x(p)) = p. \quad (2.15.1)$$

Note that from $p = f'(x(p))$ we have, according to the chain rule,

$$\frac{d}{dp} f'(x(p)) = f''(x(p)) x'(p) \implies x'(p) = [f''(x(p))]^{-1}. \quad (2.15.2)$$

From this, we can prove that $g(p)$ is itself convex:

$$\begin{aligned} g'(p) &= \frac{d}{dp} [px(p) - f(x(p))] \\ &= p x'(p) + x(p) - f'(x(p)) x'(p) = x(p), \end{aligned}$$

hence

$$g''(p) = x'(p) = [f''(x(p))]^{-1} > 0. \quad (2.15.3)$$

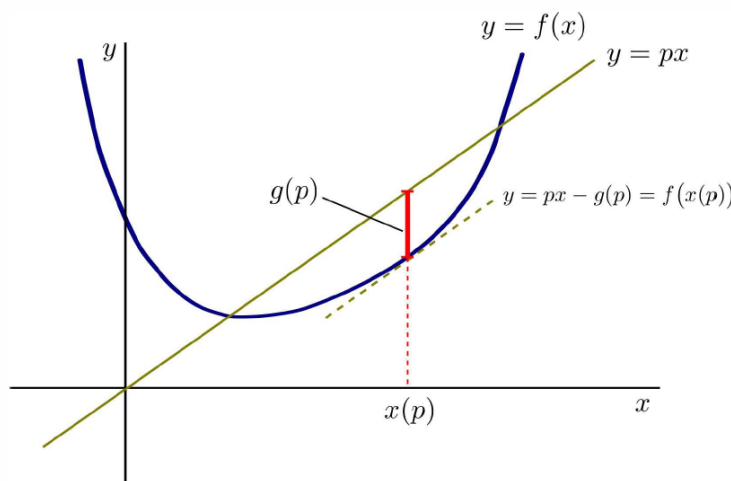


Figure 2.15.1: Construction for the Legendre transformation of a function $f(x)$.

In higher dimensions, the generalization of the definition $f''(x) > 0$ is that a function $F(x_1, \dots, x_n)$ is convex if the matrix of second derivatives, called the *Hessian*,

$$H_{ij}(\mathbf{x}) = \frac{\partial^2 F}{\partial x_i \partial x_j} \quad (2.15.4)$$

is positive definite. That is, all the eigenvalues of $H_{ij}(\mathbf{x})$ must be positive for every \mathbf{x} . We then define the Legendre transform $\mathbf{G}(\mathbf{p})$ as

$$\mathbf{G}(\mathbf{p}) = \mathbf{p} \cdot \mathbf{x} - F(\mathbf{x}) \quad (2.15.5)$$

where

$$\mathbf{p} = \nabla F. \quad (2.15.6)$$

Note that

$$dG = \mathbf{x} \cdot d\mathbf{p} + \mathbf{p} \cdot d\mathbf{x} - \nabla F \cdot d\mathbf{x} = \mathbf{x} \cdot d\mathbf{p}, \quad (2.15.7)$$

which establishes that G is a function of \mathbf{p} and that

$$\frac{\partial G}{\partial p_j} = x_j. \quad (2.15.8)$$

Note also that the Legendre transformation is *self dual*, which is to say that the Legendre transform of $G(\mathbf{p})$ is $F(\mathbf{x})$: $F \rightarrow G \rightarrow F$ under successive Legendre transformations.

We can also define a *partial Legendre transformation* as follows. Consider a function of q variables $F(\mathbf{x}, \mathbf{y})$, where $\mathbf{x} = \{x_1, \dots, x_m\}$ and $\mathbf{y} = \{y_1, \dots, y_n\}$, with $q = m + n$. Define $\mathbf{p} = \{p_1, \dots, p_m\}$, and

$$G(\mathbf{p}, \mathbf{y}) = \mathbf{p} \cdot \mathbf{x} - F(\mathbf{x}, \mathbf{y}), \quad (2.15.9)$$

where

$$p_a = \frac{\partial F}{\partial x_a} \quad (a = 1, \dots, m). \quad (2.15.10)$$

These equations are then to be inverted to yield

$$x_a = x_a(\mathbf{p}, \mathbf{y}) = \frac{\partial G}{\partial p_a}. \quad (2.15.11)$$

Note that

$$p_a = \frac{\partial F}{\partial x_a}(\mathbf{x}(\mathbf{p}, \mathbf{y}), \mathbf{y}). \quad (2.15.12)$$

Thus, from the chain rule,

$$\delta_{ab} = \frac{\partial p_a}{\partial p_b} = \frac{\partial^2 F}{\partial x_a \partial x_c} \frac{\partial x_c}{\partial p_b} = \frac{\partial^2 F}{\partial x_a \partial x_c} \frac{\partial^2 G}{\partial p_c \partial p_b}, \quad (2.15.13)$$

which says

$$\frac{\partial^2 G}{\partial p_a \partial p_b} = \frac{\partial x_a}{\partial p_b} = K_{ab}^{-1}, \quad (2.15.14)$$

where the $m \times m$ partial Hessian is

$$\frac{\partial^2 F}{\partial x_a \partial x_b} = \frac{\partial p_a}{\partial x_b} = K_{ab}. \quad (2.15.15)$$

Note that $K_{ab} = K_{ba}$ is symmetric. And with respect to the \mathbf{y} coordinates,

$$\frac{\partial^2 G}{\partial y_\mu \partial y_\nu} = -\frac{\partial^2 F}{\partial y_\mu \partial y_\nu} = -L_{\mu\nu}, \quad (2.15.16)$$

where

$$L_{\mu\nu} = \frac{\partial^2 F}{\partial y_\mu \partial y_\nu} \quad (2.15.17)$$

is the partial Hessian in the \mathbf{y} coordinates. Now it is easy to see that if the full $q \times q$ Hessian matrix H_{ij} is positive definite, then any submatrix such as K_{ab} or $L_{\mu\nu}$ must also be positive definite. In this case, the partial Legendre transform is convex in $\{p_1, \dots, p_m\}$ and concave in $\{y_1, \dots, y_n\}$.

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