1 Sufficient Conditions and Convexity (August 13)

1.1 Sufficient Convexity Condition on The Lagrangian, No Constraints

Recall that if $u_* \in \mathcal{A} = \{u \in C^1([a,b]) : u(a) = A, u(b) = B\}$ is a minimizer of the functional

$$F[u(\cdot)] := \int_a^b L(x, u(x), u'(x)) dx,$$

Then it satisfies the "primitive" Euler-Lagrange equation

$$\int_{a}^{b} \left[L_{z}(x, u_{*}(x), u'_{*}(x))v(x) + L_{p}(x, u_{*}(x), u'_{*}(x))v'(x) \right] dx = 0$$
 (*)

for all test functions v on [a, b].

Lemma 1.1. Assume the conditions above. Suppose L = L(x, z, p) is a C^1 function, and that for each $x \in [a, b]$, $L(x, \cdot, \cdot)$ is convex. If u_* satisfies (*) above, then $F[u_*(\cdot)] \leq F[u_*(\cdot) + v(\cdot)]$ for all test functions v on [a, b].

The lemma states, roughly, that if u_* is a minimizer, then it must be a global minimizer. We will look at an example of the lemma and then prove it.

(See the "model example" in the professor's notes.) Consider $L(x,z,p)=\frac{1}{2}(z^2+p^2)$. This satisfies the hypotheses of the lemma. Suppose u_* is a minimizer (of the functional defined by integrating L(x,u(x),u'(x))). Then u_* satisfies (*). The lemma then states that $F[u_*(\cdot)] \leq F[u_*(\cdot)+v(\cdot)]$ for all test functions v on [a,b]. But every function $u \in \mathcal{A}$ is of the form $u_* + v$ for some test function v on [a,b], since $u-u_*$ is a test function on [a,b]. Thus u_* is a global minimizer.

We now prove the lemma. We can think of the lemma as generalizing this model example.

Proof. Recall the C^1 criterion for convexity of a function $f: \mathbb{R}^n \to \mathbb{R}$: f is convex if and only if $f(a+b) \geq f(a) + \nabla f(a) \cdot b$ for all $a, b \in \mathbb{R}^n$. We will apply this criterion to the convex function $L(x,\cdot,\cdot): \mathbb{R} \times \mathbb{R} \to \mathbb{R}$. Directly applying the criterion gives

$$L(x,z+\widetilde{z},p+\widetilde{p}) \ge L(x,z,p) + \underbrace{\nabla_{(z,p)}L(z,z,p)\left(\widetilde{z}\atop \widetilde{p}\right)}_{=L_z(x,z,p)\widetilde{z}+L_p(x,z,p)\widetilde{p}}.$$

Then

$$F[u_*(\cdot) + v(\cdot)] = \int_a^b L(x, u_*(x) + v(x), u_*'(x) + v'(x)) dx$$

$$\geq \int_a^b L(x, u_*(x), u_*'(x)) dx + \underbrace{\int_a^b \left[L_z(\dots)v(x) + L_p(\dots)v'(x) \right] dx}_{=0}$$

$$= \int_a^b L(x, u_*(x), u_*'(x)) dx$$

$$= L[u_*(\cdot)].$$

1.2 Convex Domains

Recall that we were originally looking at convex functions $f: \Omega \to \mathbb{R}$, where f is C^1 and convex and $\Omega \subseteq \mathbb{R}^n$ is convex. We had a theorem:

Theorem 1.2. Assume the above conditions. x_* is a minimizer of f on Ω if and only if $\nabla f(x_*)(x-x_*) \geq 0$ for all $x \in \Omega$.

To this theorem there is a variational analogue. (Are we assuming the conditions of the previous subsection? This is unclear.)

Theorem 1.3. Consider a functional

$$F[u(\cdot)] := \int_a^b L(x, u(x), u'(x)) dx.$$

Let $\mathcal{A} = \{u \in C^1([a,b]) : u(a) = A, u(b) = B\}$. Let Ω be a convex subset of \mathcal{A} , and suppose that u_* is a minimizer of F on Ω . Then $F[u_*(\cdot) + sv(\cdot)] \geq F[u_*(\cdot)]$ for all s and all test functions v on [a,b], and

$$\int_{a}^{b} \frac{\delta F}{\delta u}(u_{*})(x)(u(x) - u_{*}(x)) dx \int_{a}^{b} \frac{\delta F}{\delta u}(u_{*})(x)v(x) dx = \frac{d}{ds} \bigg|_{s=0} F[u_{*}(\cdot) + sv(\cdot)] \ge 0$$

for all $u \in \Omega$. (See how the above condition parallels to the finite-dimensional case.)

We will not prove this, but we will look at an example. Suppose

$$F[u(\cdot)] = \int_0^1 \frac{1}{2} \left(u'(t)^2 + u(t) \right) dt,$$

where \mathcal{A} is the set of C^1 functions on [0,1] that are zero at the endpoints, and $\Omega = \{u \in \mathcal{A} : u(t) \ge \sin^2(\pi t)\}$. One can check that Ω is convex. The Lagrangian function is $L(x,z,p) = \frac{1}{2}(p^2+z)$. This is a convex function of (z,p), so it satisfies the conditions of the first lemma. Then if u_* minimizes F on Ω , let $f(s) = F[(1-s)u_*(\cdot) + su(\cdot)]$. We are asked to show that $f'(0) \ge 0$. (Author could not write this part.)