

Muskat问题的极值原理

Maximum principle for Muskat problem

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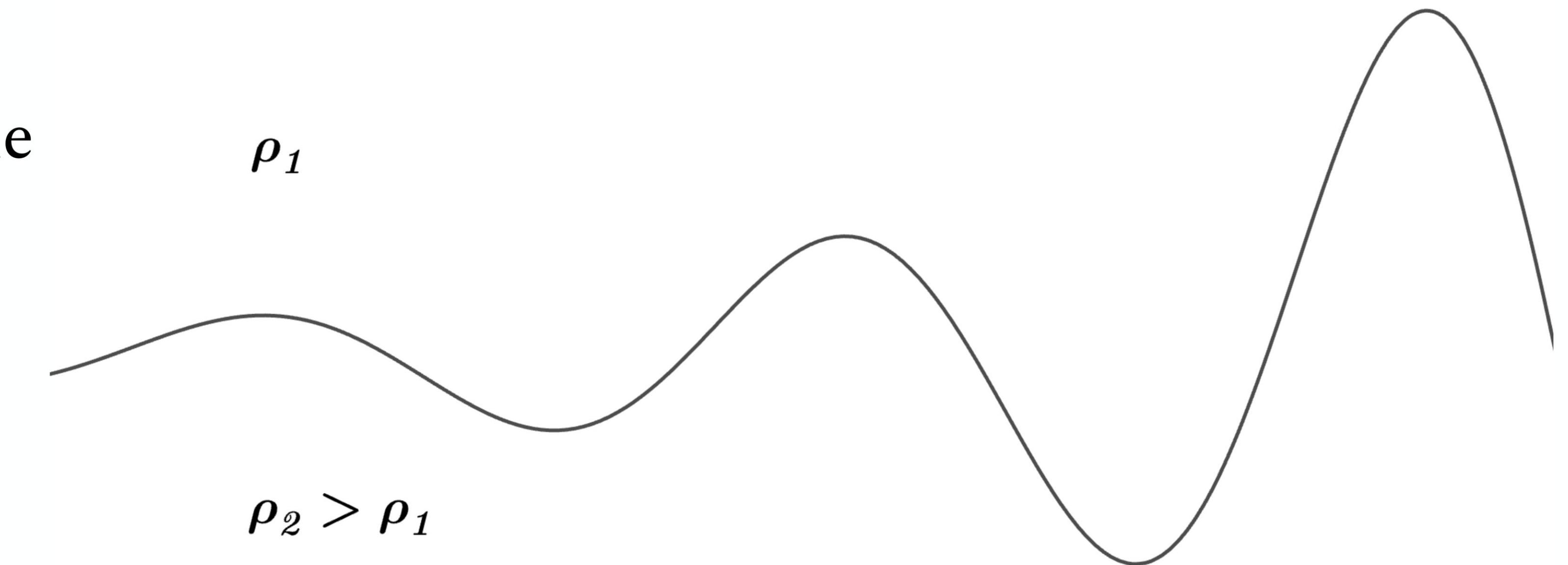
Introduction to Muskat Problem

Muskat問題

- free boundary problem
- the motion of the interface between 2 incompressible, immiscible fluids of different constant densities in a porous media

Global regularity vs finite-time singularity:

- Flatten over time?
- Finite-time turning of the interface?



Muskat問題

- the equation for the interface:

$$\partial_t f(t, x) = \int_{\mathbb{R}} \frac{(f(t, y) - f(t, x)) - (y - x)f_x(t, x)}{(f(t, y) - f(t, x))^2 + (y - x)^2} dy$$

- linearized equation:

$$\partial_t f(t, x) = \int_{\mathbb{R}} \frac{f(t, y) - f(t, x)}{(y - x)^2} dy$$

$$\partial_t f(t, x) = - C(-\Delta)^{\frac{1}{2}} f(t, x)$$

- Parabolic equation.

Muskat问题的已有研究成果

- [Córdoba and Gancedo 2009]: L^∞ maximum principle:

$$\| f(t, \cdot) \|_{L^\infty} \leq \| f(0, \cdot) \|_{L^\infty}.$$

- [Córdoba and Gancedo 2009]: Local well-posedness in H^k for $k \geq 3$.
- [Constantin et al. 2013]: L^2 maximum principle:

$$\| f(t, \cdot) \|_{L^2} \leq \| f(0, \cdot) \|_{L^2}.$$

Muskat问题的尺度变换性质

- Scaling property: If $f(t, x)$ is a solution with initial data $f_0(x)$,

$$f_\lambda(t, x) := \lambda f\left(\frac{t}{\lambda}, \frac{x}{\lambda}\right)$$

is a solution with initial data $f_{\lambda,0}(x) = \lambda f_0\left(\frac{x}{\lambda}\right)$.

- Scaling invariant norm, or critical norm: Norm $\|\cdot\|_X$ such that

$$\|f_0\|_X = \|f_{\lambda,0}\|_X, \quad \forall f_0.$$

- Examples: $\|f\|_{\dot{W}^{1,\infty}} := \|f_x\|_{L^\infty}$ and $\|f\|_1 := \|\hat{\xi f}(\xi)\|_{L^1}$.

Muskat问题的已有研究成果

Well-posedness theory for scaling invariant norms:

- [Constantin et al. 2013]: Exists global weak Lipschitz solution when $\|f'_0\|_{L^\infty} < 1$.
- [Constantin et al. 2016]: Globally well-posed when $\|f_0\|_1 = \||\xi|\hat{f}_0(\xi)\|_{L_\xi^1} \lesssim \frac{1}{3}$.
- [Constantin et al. 2017]: H^k initial data stays H^k , provided that $f_x(t, \cdot)$ is bounded and has some modulus of continuity.
- [Cameron 2019]: Maximum principle and global well-posedness when $\|f'_0\|_{L^\infty} < 1$.

Muskat问题的已有研究成果

What happens when $\|f'_0\|_{L^\infty} > 1$?

- Loss of parabolicity \implies finite time blow-up(?)

My results: A priori estimates of scaling invariant norm $\|f_x(t, \cdot)\|_{L^\infty}$ when $\|f'_0\|_{L^\infty} > 1$.

- Maximum principle of $\|f_x(t, \cdot)\|_{L^\infty}$ when $\|f'_0\|_{L^\infty} \leq L \approx 2.98$,
- Optimality of the constant L ,
- Power law decay of $\|f_x(t, \cdot)\|_{L^\infty}$ when $\|f'_0\|_{L^\infty} < L$.

Muskat问题的极值原理

Theorem I Suppose $f(t, x) \in C_t^1([0, T], C_x^1(\mathbb{R}))$ is a smooth solution with initial data $f_0(x)$ such that $\|f'_0\|_{L_x^\infty} \leq L$, where $L \approx 2.98$ is the first positive root of the following trigonometric equation

$$\begin{cases} 2 \cos(\alpha) + 3 \cos\left(\frac{5\alpha}{3}\right) - \cos(3\alpha) = 0, \\ L = \tan(\alpha). \end{cases}$$

Then, $f_x(t, x)$ has maximum principle. To wit, $\|f_x(t, \cdot)\|_{L_x^\infty}$ decreases in time.

Remark on theorem I: The constant $L \approx 2.98$ is the optimal constant for maximum principle to be hold. To wit, for any $\varepsilon > 0$, there exists a solution $f_\varepsilon(t, x)$, such that $\|f'_{0,\varepsilon}(0, \cdot)\|_{L_x^\infty} = L + \varepsilon$, and $\frac{d}{dt} \|f_\varepsilon(t, \cdot)\|_{L_x^\infty} \Big|_{t=0} > 0$.

Muskat问题的极值原理

Theorem II Suppose $f(t, x) \in C_t^1([0, T], C_x^1(\mathbb{R}))$ is a smooth solution with initial data $f_0(x)$ such that $\|f'_0\|_{L_x^\infty} < L$, where $L \approx 2.98$ is defined as in theorem I. Then, $\|f_x(t, \cdot)\|_{L_x^\infty}$ decays at a power law rate in time. To be specific,

$$\begin{cases} \|f_x(t, \cdot)\|_{L_x^\infty} \leq \frac{C \|f_0\|_{L^\infty}}{t}, \\ \|f_x(t, \cdot)\|_{L_x^\infty} \leq \frac{C \|f_0\|_{L^2}}{t^{3/2}}, \end{cases}$$

where C is a constant that depends only on $\|f'_0\|_{L^\infty}$.

Remark on theorem II: The power law decay rate is optimal.

Muskat問題的極值原理

- Main difficulty: the loss of parabolicity when $\|f'_0\|_{L^\infty} > 1$.
- the equation of $f_x(t, x)$:

$$\begin{aligned}\partial_t f_x(t, x) = & -\partial_x f_x(t, x) \cdot \int_{\mathbb{R}} \frac{y - x}{(f(t, y) - f(t, x))^2 + (y - x)^2} dy \\ & - 2 \int_{\mathbb{R}} \frac{f_x(t, x) - l(t, y, x)}{(y - x)^2} \cdot \frac{1 + f_x(t, x)l(t, y, x)}{\left(1 + l^2(t, y, x)\right)^2} dy,\end{aligned}$$

where $l(t, y, x) := \frac{f(t, y) - f(t, x)}{y - x}$ is the slope between $(x, f(t, x))$ and $(y, f(t, y))$.

- If $\|f'_0\|_{L^\infty} < 1$, $1 + f_x(t, x)l(t, y, x)$ is positive $\implies \partial_t f_x(t, x) \approx \text{drift term} - C(-\Delta)^{\frac{1}{2}} f_x(t, x)$.
- If $\|f'_0\|_{L^\infty} > 1$, $1 + f_x(t, x)l(t, y, x)$ is indefinite \implies loss of parabolicity.

Motivation

证明的动机

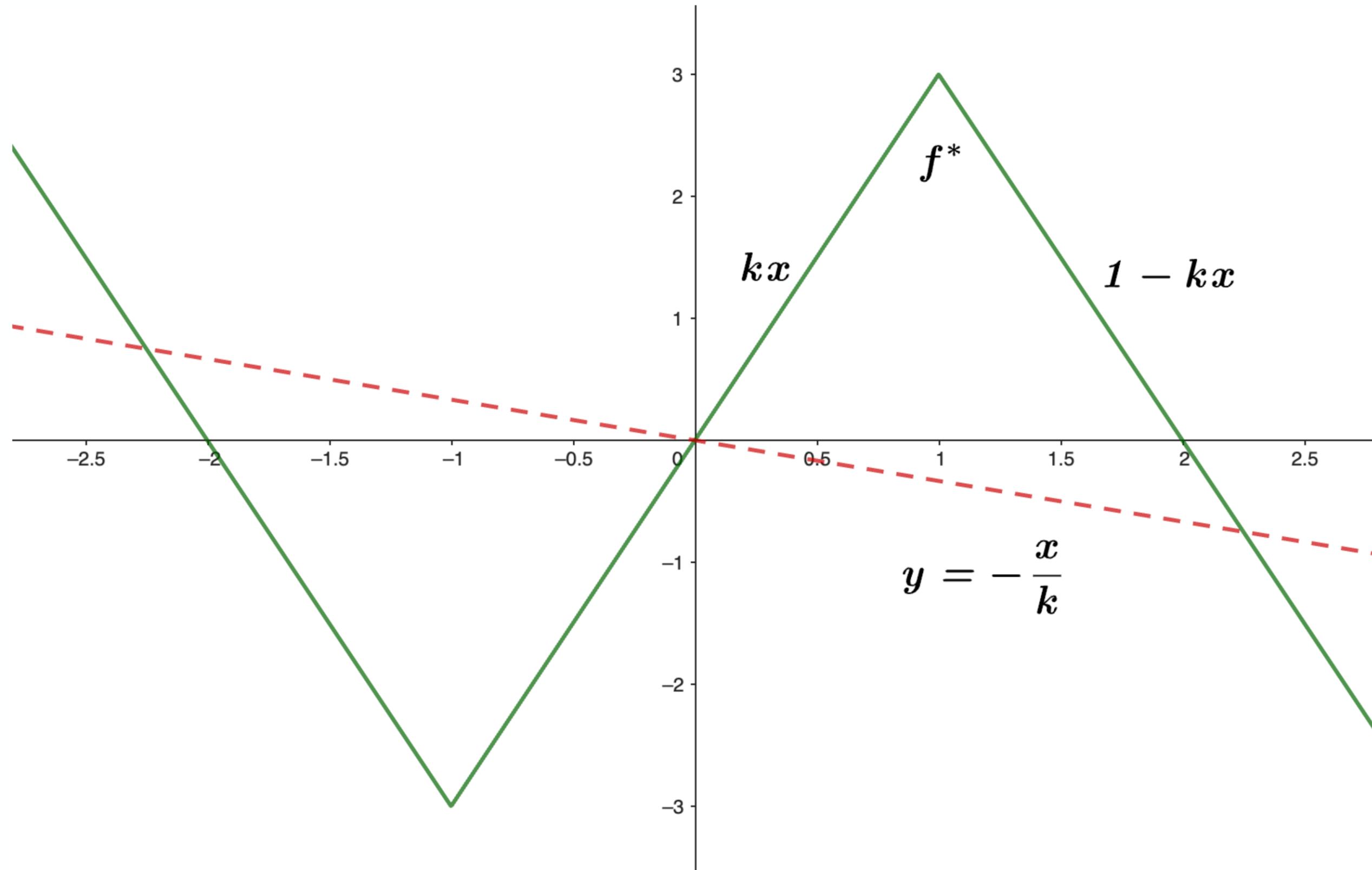
- Assuming $f_x(t, x^*) = \max_{x \in \mathbb{R}} |f_x(t, x)|$, we want to show $\partial_t f_x(t, x^*) \leq 0$.

$$\begin{aligned}\partial_t f_x(t, x^*) &= -\partial_x f_x(t, x^*) \cdot \int_{\mathbb{R}} \frac{y - x^*}{(f(t, y) - f(t, x^*))^2 + (y - x^*)^2} dy \\ &\quad - 2 \int_{\mathbb{R}} \frac{f_x(t, x^*) - l(t, y, x^*)}{(y - x^*)^2} \cdot \frac{1 + f_x(t, x^*)l(t, y, x^*)}{(1 + l^2(t, y, x^*))^2} dy,\end{aligned}$$

where $l(t, y, x^*) := \frac{f(t, y) - f(t, x^*)}{y - x^*}$ is the slope between $(x^*, f(t, x^*))$ and $(y, f(t, y))$.

- $f_x(t, x^*) - l(t, y, x^*) > 0$, and $1 + f_x(t, x^*)l(t, y, x^*) > 0$ if $l(t, y, x^*) > -\frac{1}{f_x(t, x^*)}$.
- Integrand < 0 only if $l(t, y, x^*) < -\frac{1}{f_x(t, x^*)}$.

设想中的最坏情况



conjectured worst case scenario

- $l(t, y, 0)$ reaches $-\frac{1}{f_x(t, 0)}$ as fast as possible in the conjectured worst case.
- $\frac{d}{dt} \|f_x(t, \cdot)\|_{L^\infty} \approx -\frac{k(3 - k^2)}{(1 + k^2)^2} \leq 0$, provided that $k = \|f_x\|_{L^\infty} \leq \sqrt{3}$.
- Maximum principle holds when $\|f'_0\| \lesssim \sqrt{3}$?

The Key Idea of the Proof

极值原理的证明: 命题的转化

- Define

$\mathcal{F}_k := \{f \in \mathbb{R}^{[0,+\infty)} : \text{piecewise differentiable}, f_x(0) = k, -k \leq f_x(x) \leq k\},$

$$H_k(l) := \frac{2(k-l)(1+kl)}{(1+l^2)^2},$$

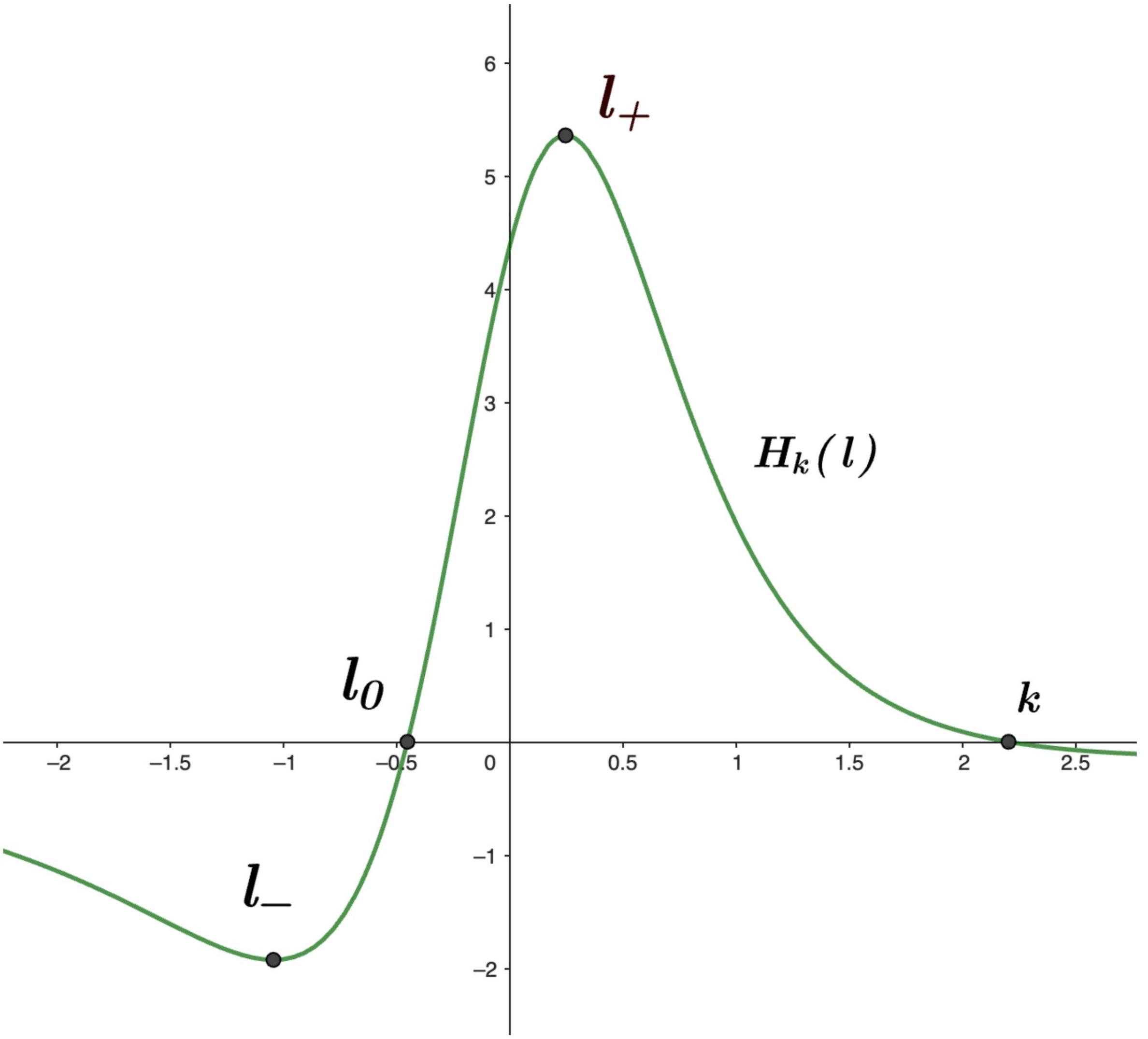
$$I_{[f]} := \int_0^\infty \frac{2}{y^2} \cdot \frac{\left(f_x(0) - l_{[f]}(y)\right) \cdot \left(1 + f_x(0)l_{[f]}(y)\right)}{\left(1 + l_{[f]}^2(y)\right)^2} dy = \int_0^\infty \frac{1}{y^2} H_k(l_{[f]}(y)) dy,$$

where $l_{[f]}(y) := \frac{f(y) - f(0)}{y}$.

- Need to prove: If $k \leq L$, $I_{[f]} \geq 0$ for any $f \in \mathcal{F}_k$.
- Problem of calculus of variations. What is the minimizer of the functional $I_{[f]}$?

极值原理的证明: 积分核的单调性

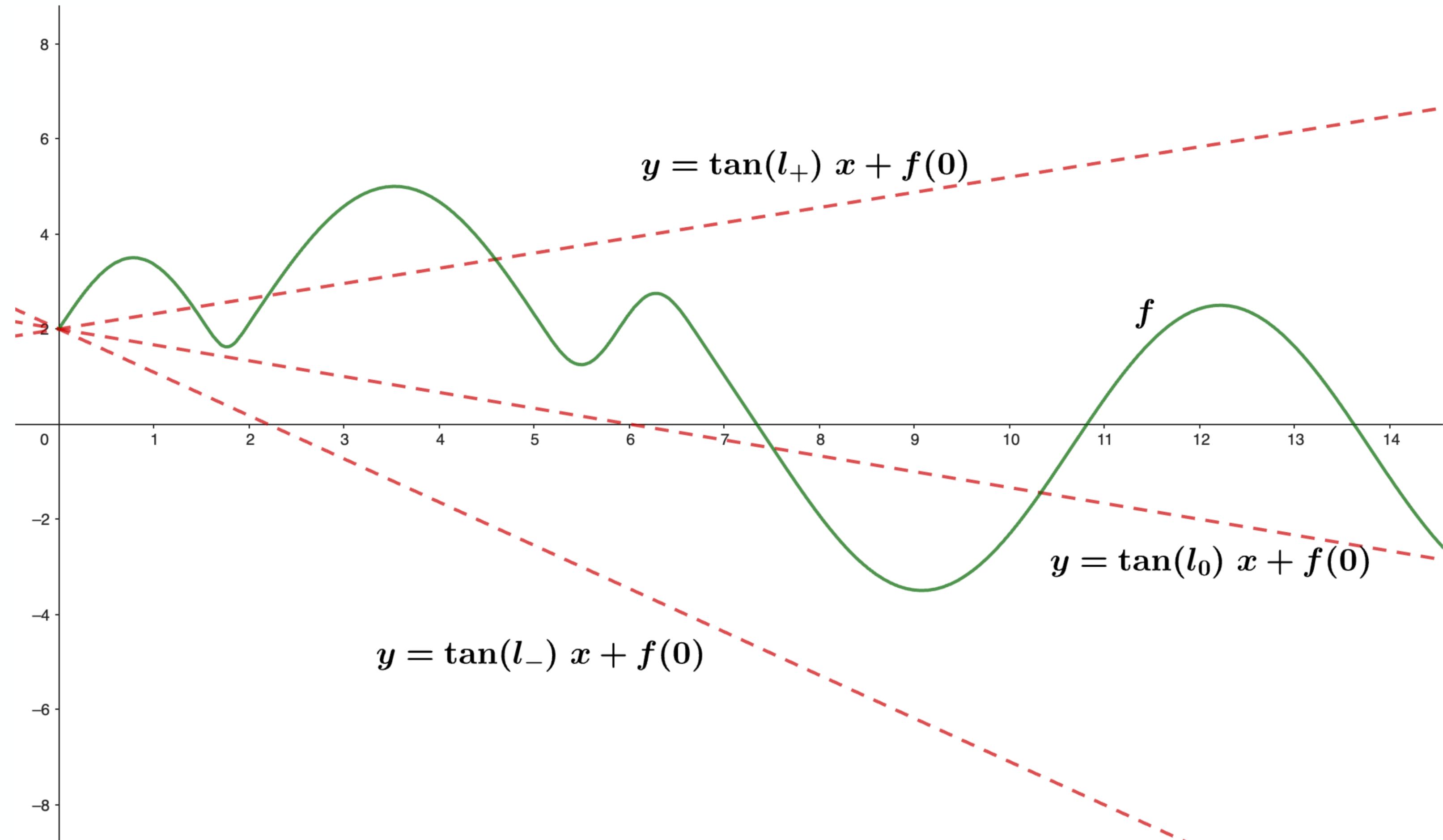
- $I_{[f]} = \int_{\mathbb{R}} \frac{1}{y^2} H_k(l_{[f]}(y)) dy$, where
 $l_{[f]}(y)$ is the slope and $k = \|f_x\|_{L^\infty}$.
 - $H_k(l) := \frac{2(k - l)(1 + kl)}{(1 + l^2)^2}$.
- $\begin{cases} H_k(l) \text{ decreases ,} & l \in (-\infty, l_-], \\ H_k(l) \text{ increases ,} & l \in [l_-, l_+], \\ H_k(l) \text{ decreases ,} & \beta \in [l_+, k], \\ H_k(l_-) = \min_{-\infty \leq l \leq k} H_k(l), & \\ H_k(l) \geq 0, & \beta \in [l_0, k], \\ H_k(l) \leq 0, & \beta \in (-\infty, l_0]. \end{cases}$



极值原理的证明: 最小化过程

- $I_{[f]} = \int_{\mathbb{R}} \frac{1}{y^2} H_k(l_{[f]}(y)) dy,$

where $l_{[f]}(y)$ is the slope between $(y, f(y))$ and $(0, f(0))$, and $k = \|f_x\|_{L^\infty}$.

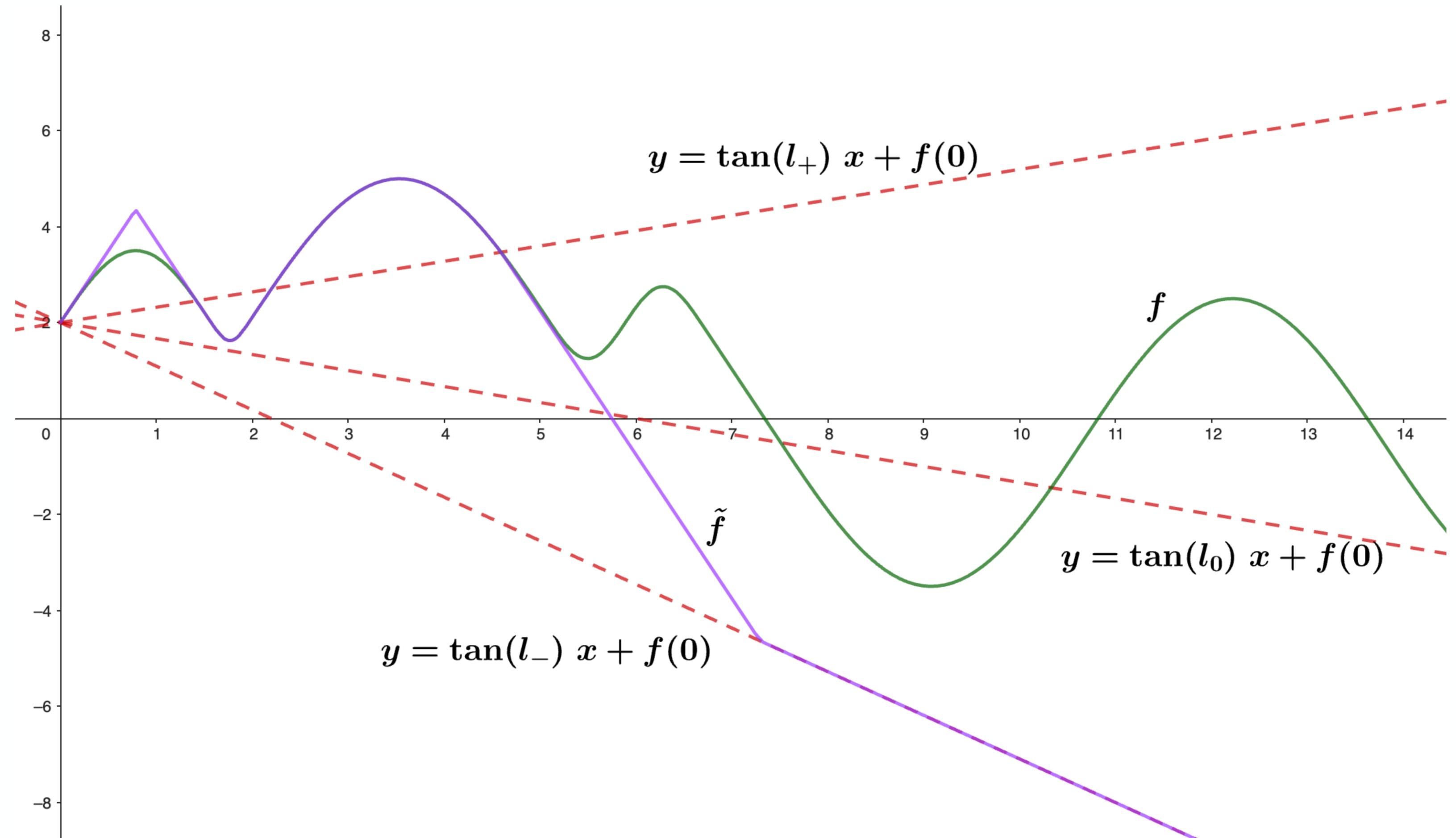


极值原理的证明: 最小化过程

- $I_{[f]} = \int_{\mathbb{R}} \frac{1}{y^2} H_k(l_{[f]}(y)) dy,$

where $l_{[f]}(y)$ is the slope between $(y, f(y))$ and $(0, f(0))$, and $k = \|f_x\|_{L^\infty}$.

- $H_k(l_{[\tilde{f}]}(y)) \leq H_k(l_{[f]}(y)).$
- $I_{[\tilde{f}]} \leq I_{[f]}.$

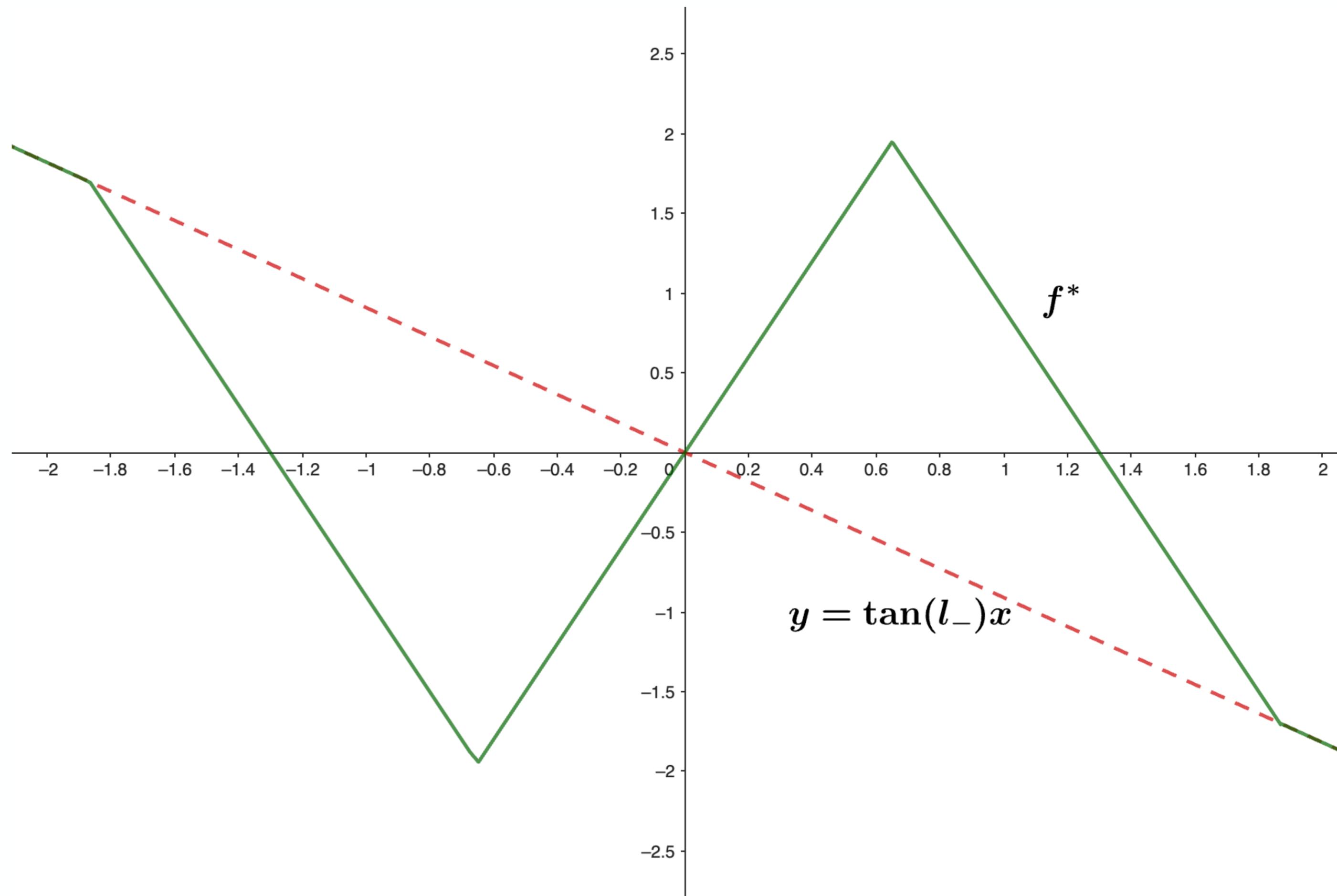


极值原理的证明: 最小化子的计算

$$I_{[\tilde{f}]} \geq \begin{cases} \frac{1}{x_+(k + \tan(\frac{2\alpha}{3} - \frac{\pi}{6}))} \left[1 - F\left(\tan(\frac{2\alpha}{3} - \frac{\pi}{6})\right) \right] \\ \quad + \frac{1}{x_2(k + \tan(\frac{2\alpha}{3} - \frac{\pi}{6}))} \left[F\left(\tan(\frac{2\alpha}{3} - \frac{\pi}{6})\right) - F(-k) \right], & \alpha \leq \frac{3\pi}{10}, \\ \frac{1}{x_+(k + \tan(\frac{2\alpha}{3} - \frac{\pi}{6}))} \left[1 - F\left(\tan(\frac{2\alpha}{3} - \frac{\pi}{6})\right) \right] \\ \quad + \frac{1}{x_2(k + \tan(\frac{2\alpha}{3} - \frac{\pi}{6}))} \left[F\left(\tan(\frac{2\alpha}{3} - \frac{\pi}{6})\right) - F\left(\tan(\frac{2\alpha}{3} - \frac{\pi}{2})\right) \right] \\ \quad + \frac{k + \tan(\frac{2\alpha}{3} - \frac{\pi}{2})}{x_2(k + \tan(\frac{2\alpha}{3} - \frac{\pi}{6}))} H_k(\tan(\frac{2\alpha}{3} - \frac{\pi}{2})), & \alpha > \frac{3\pi}{10}, \end{cases}$$

where $\alpha := \arctan k$ and $F(l) := \frac{2(k-l)(1+kl)}{1+l^2} \implies I_{[\tilde{f}]} \geq 0$ if $k = \|f'_0\|_{L^\infty} \leq L$.

极值原理的证明: L 的最优化



worst case scenario

衰减性的证明

Define $M(t) := \max_{x \in \mathbb{R}} |f_x(t, x)|$, then

$$\begin{aligned} M'(t) &= - \int_{\mathbb{R}} \frac{2}{(y - x_t)^2} \cdot \frac{(f_x(t, x_t) - l(t, y, x_t)) \cdot (1 + f_x(t, x_t)l(t, y, x_t))}{(1 + l^2(t, y, x_t))^2} dy \\ &= - \int_{\mathbb{R}} \frac{2}{(y - x_t)^2} \frac{\varepsilon (f_x(t, x_t) - l(t, y, x_t))}{(1 + l^2(t, y, x_t))^2} dy \\ &\quad - \int_{\mathbb{R}} \frac{2}{(y - x_t)^2} \cdot \frac{(f_x(t, x_t) - l(t, y, x_t)) \cdot (1 - \varepsilon + f_x(t, x_t)l(t, y, x_t))}{(1 + l^2(t, y, x_t))^2} dy \\ &=: -\varepsilon I_1 - I_{2,\varepsilon}, \end{aligned}$$

where x_t is a maximum point.

衰减性的证明

- $M(t) := \max_{x \in \mathbb{R}} |f_x(t, x)|,$

$$\begin{aligned} M'(t) &= - \int_{\mathbb{R}} \frac{2}{(y - x_t)^2} \frac{\varepsilon (f_x(t, x_t) - l(t, y, x_t))}{(1 + l^2(t, y, x_t))^2} dy \\ &\quad - \int_{\mathbb{R}} \frac{2}{(y - x_t)^2} \cdot \frac{(f_x(t, x_t) - l(t, y, x_t)) \cdot (1 - \varepsilon + f_x(t, x_t)l(t, y, x_t))}{(1 + l^2(t, y, x_t))^2} dy \\ &=: -\varepsilon I_1 - I_{2,\varepsilon}. \end{aligned}$$

- $I_1 \approx C(-\Delta)^{\frac{1}{2}} f_x(t, x_t)$
- $I_{2,\varepsilon}$ is a perturbation of $I_{[f]}$, thus $I_{2,\varepsilon} \geq 0$ for some small enough $\varepsilon > 0$.
- $M'(t) \leq -\varepsilon C(-\Delta)^{\frac{1}{2}} f_x(t, x).$

Thanks