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J. Frederic Bonnans

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Optimization of running strategies based on anaerobic energy and variations of velocity

J. Frédéric Bonnans

Inria-Saclay and CMAP, Ecole Polytechnique, France

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Outline

Keller's model

2 Variable energy recreation

 \blacksquare Bounding the derivative of f

Sources

- Talk based on the joint work [1] with Amandine Aftalion, Lab. Math. Versailles, UVSQ.
- Pioneering reference: J.B. Keller [2].
- [1] A. Aftalion and J.F. Bonnans, *Optimization of running strategies based on anaerobic energy and variations of velocity*. SIAM J. Applied Math., to appear (HAL preprint).
- [2] Keller, J.B., *Optimal velocity in a race*, Amer. Math. Monthly 81 (1974), 474–480.

1 Keller's model

2 Variable energy recreation

 \blacksquare Bounding the derivative of f

Keller's model: dynamics

Consider the following **state equation**:

$$\dot{v}(t) = f(t) - \phi(v(t)); \quad \dot{e}(t) = \bar{\sigma} - f(t)v(t),$$

- **Cost function**: $-\int_0^T v(t) dt$.
- **energy recreation**: $\bar{\sigma} > 0$
- drag function $\phi(v) = v/\tau$, with $\tau > 0$.
- **Generalized drag function**: ϕ C^2 on $(0, \infty)$; $\phi(0) = 0$; ϕ' positive; $v\phi'(v)$ increasing **Example**: $\phi(v) = av^{\beta}$, a > 0, $\beta > 1$.

Keller's model: constraints

- Control constraint: $0 \le f(t) \le f_M$.
- Energy constraint: $e(t) \ge 0$.
- Maximal force strategy: optimal for short horizons $T \leq T_c$.
- In the sequel $T > T_c$.

Keller's model: discussion

■ Maximal speed (never reached): $0 = f_M - v_M/\tau$, i.e.

$$v_{M}=\tau f_{M}.$$

■ Constant energy speed: $f = v/\tau$, $\sigma = fv$, and so:

$$v_E = \sqrt{\tau \sigma}$$
.

Energy variation when constant speed:

$$e(t) \downarrow \text{ if } v > v_E$$
, and $e(t) \uparrow \text{ otherwise}$.

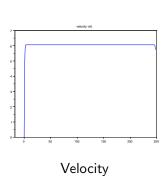
Keller's model: Keller's conjecture

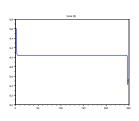
- Optimal strategy has three arcs
- First arc: **maximal force** $f = f_M$.
- Second arc: **constant speed** $v > v_F$.
- Last arc: **zero energy** e = 0.
- Negative jump of force at junction times
- Unclear proofs in the literature for $\phi(v) = v/\tau$.

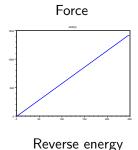
Keller's model: numerical results!

- Mechanical parameters: $f_{max} = 9.6 m \ s^{-2}$, $\tau = 0.67 s$
- Energy parameters: $e_{an}^0 = 1400 m^2 s^{-3}$, $\bar{\sigma} = 49 m^2 s^{-3}$.
- d = 1500m.
- 2000 discretization steps, i.e. time step close to 0.12s.
- Free software www.bocop.org.
- Optimal time is 248.21s

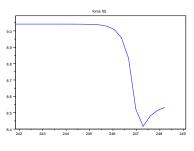
Keller's model: numerical results with bocop.org



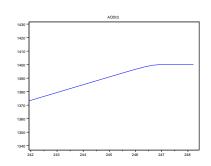




Zoom on end of race: zero energy arc



Final force



Final reverse energy

Keller's model: main result

So the numerical results agree with Keller's conjecture, and indeed:

Theorem

Keller's conjecture holds, even with the generalized drag function.

We next present the proof of this result.

Keller's model: Hamiltonian and costate

Hamiltonian function:

$$H:=-v+p_{v}(f-\phi(v))+p_{e}(\bar{\sigma}-fv).$$

Costate equation:

$$\begin{cases} -\dot{p}_{v} &= -1 - p_{v} \phi'(v) - p_{e} f \\ -\mathrm{d} p_{e} &= -\mathrm{d} \mu, \end{cases}$$

Final conditions $p_v(T) = p_e(T) = 0$, and so: $p_e = \mu$. Multiplier to the state constraint: say $\mu(T) = 0$ and

$$d\mu \geq 0$$
; $\sup(d\mu) \subset \{t; e(t) = 0\}.$

Therefore $p_e = \mu < 0$.



Switching function

The switching function is

$$\Psi := H_u = p_v - p_e v.$$

By Pontryagin's principle, we have:

$$f(t) = \left\{ egin{array}{ll} f_M & ext{if } \Psi(t) > 0, \\ 0 & ext{if } \Psi(t) < 0, \end{array}
ight.$$

p_e has negative values

Lemma

pe has negative values.

Proof.

If $p_e(\tau) = 0$ then over (τ, T) :

 $p_e = 0$, since it is nondecreasing and $p_e(T) = 0$.

$$\dot{p}_{
m v}=1+p_{
m v}\phi'({
m v})$$
 (is >0 if $p_{
m v}\geq 0$) and $p_{
m v}({
m \it T})=0$

So
$$p_{v} < 0 \Rightarrow \Psi = p_{v} < 0 \Rightarrow f = f_{M}$$

Contradiction by lemma 3.7 of paper.



p_{ν} has negative values

Lemma

 p_v has negative values.

Proof.

$$p_{\nu} \ge 0 \Rightarrow \Psi > 0 \Rightarrow f = 0 \Rightarrow \dot{p}_{\nu} > 0; \text{ but } p_{\nu}(T) = 0.$$

Corollary

An optimal trajectory starts with a maximal force arc.

Derivative of switching function

When the state constraint is not active, p_e is constant and so:

$$\Psi := H_u = p_v - p_e v.$$

$$\dot{\Psi} = (1 + p_{\nu}\phi'(\nu) + p_{e}f) - p_{e}(f - \phi(\nu))$$

= $1 + p_{\nu}\phi'(\nu) + p_{e}\phi(\nu)$

and so,

$$\dot{\Psi} - \Psi \phi'(v) = 1 + p_{\mathsf{e}}(\phi(v) + v\phi'(v)).$$

Lemma

On a singular arc, v is constant.

No zero force arc

$$\Delta := \dot{\Psi} - \Psi \phi'(v) = 1 + p_e(\phi(v) + v\phi'(v)).$$

Lemma

No zero force arc occurs.

- Let (t_a, t_b) be such an arc; then $t_a > 0$.
- If $t_b = T$ then e(T) > 0: impossible.
- $\Psi \geq 0$ over the arc; $\Psi(t_a) = \Psi(t_b) = 0$.
- We have therefore $\Delta(t_a) \geq 0 \geq \Delta(t_b)$.
- On this arc p_e is constant and v decreases, and so $\Delta(t)$ increases: contradiction.



No second maximal force arc

$$\Delta := \dot{\Psi} - \Psi \phi'(v) = 1 + p_{\mathsf{e}}(\phi(v) + v \phi'(v)).$$

Lemma

No second maximal force arc occurs.

- Let (t_a, t_b) be such an arc; then $t_a > 0$.
- If $t_b < T$, "symmetric argument": v increases, Δ decreases, but $\Delta(t_a) \le 0 \le \Delta(t_b)$.
- If $t_b = T$, μ should have a jump at time T, but then

$$\lim_{t\uparrow T} \Psi(t) = -p_e(T_-)v(T) > 0$$

implying that f = 0 at the end of the trajectory.



Keller's model: proof of main result

- Let $t_a \in (0, T)$ be the exit point of the maximal force arc.
- We know that $\Psi = 0$ over (t_a, T) .
- Let $t_b \in (0, T)$ be the first time at which the energy vanishes $(\mu \text{ has no final jump}).$
- \bullet (t_a, t_b) is a singular arc.
- If the energy is not zero on (t_b, T) : there would exist t_c, t_d with $t_b \le t_c < t_d \le T$ such that $e(t_c) = e(t_d) = 0$, and e(t) > 0, for all $t \in (t_c, t_d)$.
- Then (t_c, t_d) is a singular arc, over which $\dot{e} = \sigma fv$ is constant: contradiction.

1 Keller's model

2 Variable energy recreation

3 Bounding the derivative of *f*

Variable energy recreation

Less recreation when energy close to its extreme values, and additional recreation when deceleration:

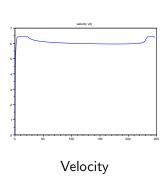
$$\dot{e} = \sigma(e) + \eta(a) - fv$$

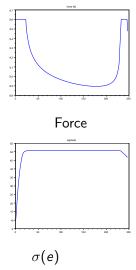
where a is the acceleration: $a = \dot{v} = f - \phi(v)$, and η convex:

$$\eta(a) \ge 0, \quad \eta(a) = 0 \text{ iff } a \ge 0.$$

■ Function $\sigma(e)$ regularization of a piecewise affine and continuous function, value 0 at extreme values, otherwise constant.

Numerical results with variable $\sigma(e)$ and $\eta(a) = 0$





Hamiltonian and costate with variable $\sigma(e)$

Hamiltonian function:

$$H:=-v+p_{v}(f-\phi(v))+p_{e}(\sigma(e)+\eta(f-\phi(v))-fv).$$

Costate equation:

$$\begin{cases} -\dot{p}_{v} &= -1 - p_{v}\phi'(v) - p_{e}f - p_{e}\eta'(a)\phi'(v) \\ -\mathrm{d}p_{e} &= p_{e}\sigma'(e) - \mathrm{d}\mu, \end{cases}$$

Final conditions
$$p_v(T) = p_e(T) = 0$$
.
 $\mu(T) = 0$, $d\mu \ge 0$, $\operatorname{supp}(d\mu) \subset \{t; e(t) = 0\}$.

Minimization of Hamiltonian with variable $\sigma(e)$

■ Hamiltonian function:

$$H:=(p_v-p_ev)f+p_e\eta(f-\phi(v))+$$
 indep. of f

- **Concave function** of the control *f*
- Minimum attained at extreme values 0 or f_M .
- Need of relaxed formulation; expectation of force:

$$f = 0 \times (1 - \theta) + \theta f_M$$

■ We may as well take f as parameter and then $\theta = f/f_M$.

Relaxed formulation

Recreation due to deceleration at zero speed:

$$R(v) := \eta(-\phi(v))$$

State equation

$$\dot{v} = f - \phi(v)
\dot{e} = \sigma(e) - fv + (1 - f/f_M)R(v)$$

Hamiltonian

$$H = p_{v}(f - \phi(v)) + p_{e}(\sigma(e) - fv + (1 - f/f_{M})R(v)).$$

Relaxed formulation: theoretical analysis

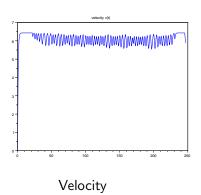
- Assume $\sigma(e) = \bar{\sigma}$ and $\eta(a) = \varepsilon \eta'(a)$
- For $\varepsilon > 0$ small enough
- Same optimal structure as for Keller'model: maximal force, constant speed, zero energy.

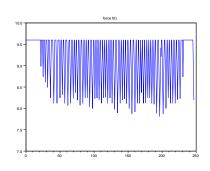
1 Keller's model

2 Variable energy recreation

 \blacksquare Bounding the derivative of f

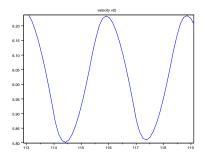
Numerical results when bounding \dot{f}



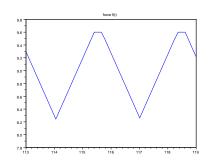


Force

Numerical results: zoom with variable $\sigma(e)$ and $\eta(a) \neq 0$



Zoom on velocity



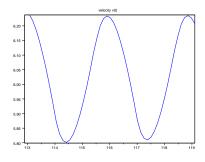
Zoom on force

Periodic problem

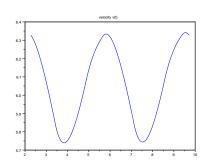
- Maximize average speed: $(1/T) \int_0^T v(t) dt$.
- Periodic speed and force, loss of energy $e(0) = e(T) + Te_d$.
- State equation and constraints

$$\begin{cases} v(0) = v(T); & f(0) = f(T). \\ e(0) = e(T) + Te_d; \\ \dot{v} = f - v\tau; \\ \dot{e} = \sigma + \eta(a) - fv; \\ 0 \le f \le f_M; & |\dot{f}| \le 1, \text{ for a.e. } t \in (0, T). \end{cases}$$

Zoom vs periodic: velocity

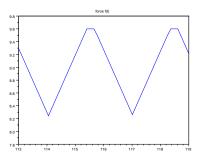


Zoom on velocity

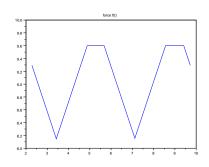


Periodic velocity

Zoom vs periodic: force



Zoom on force



Periodic force

Related work

[1] S. Aronna, J.F. Bonnans and B.S. Goh, Second order necessary conditions for control-affine problems with state constraints. Research report, to appear (HAL preprint).

Open questions

- Asymptotic analysis $\eta(a) = \varepsilon \eta'(a)$.
- Shape of oscillations ?
- Expansion of cost function

THE END!