

# Extending the JKO scheme beyond gradient flows

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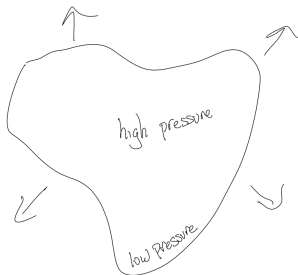
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# Tumor growth models

$$\begin{cases} \partial_t \rho - \nabla \cdot (\rho \nabla p) = \rho G(\rho), \\ p = e'(\rho) \end{cases} \quad (1)$$

- Models the growth of a tumor (or cell population) where the main limitation to growth is a competition for space.
- $p = e'(\rho)$  is the system pressure where  $e$  is a convex and increasing function.
- $G$  is a decreasing function determining the growth rate.



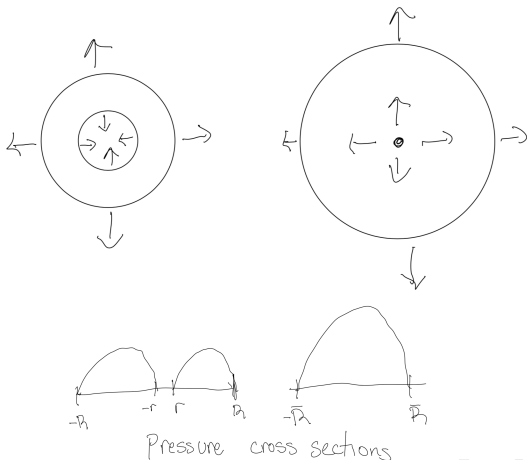
- A common choice for the pressure is  $p = \rho^\gamma$  for some  $\gamma > 0$ .
- Sending  $\gamma \rightarrow \infty$ , the pressure becomes a Lagrange multiplier for the incompressibility constraint  $\rho \leq 1$ .

$$\begin{cases} \partial_t \rho - \nabla \cdot (\rho \nabla p) = \rho G(p), \\ \rho(1 - \rho) = 0 \end{cases} \quad (2)$$

- This is a free boundary problem with a sharp interface between the occupied/empty regions i.e.  $\rho \in \{0, 1\}$ .

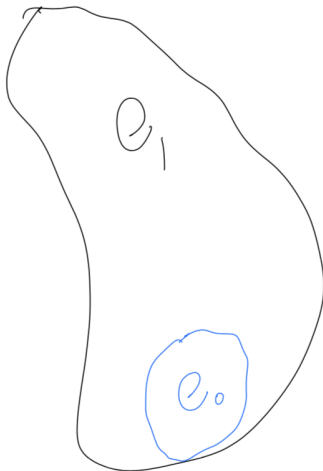
# Challenges/Features

- The tumor may undergo topological changes as it grows.
- Pressure regularity can badly degenerate at topological changes.



# Challenges/Features

- The equation satisfies a comparison principle.
- Not entirely obvious since smaller mass has a lower pressure and hence faster growth rate.



Can we design a numerical method to simulate this model that satisfies the following properties?

- Unconditionally stable.
- Preserves the comparison principle.
- Converges to the continuum PDE as the approximation vanishes.

Very tempting to use a JKO (like) scheme!

$$\begin{cases} \partial_t \rho - \nabla \cdot (\rho \nabla \rho) = \rho G(\rho), \\ \rho = e'(\rho) \end{cases}$$

- View the equation as a gradient flow with respect to unnormalized/unbalanced optimal transport (Chizat, Di Marino 2018).
- Use a splitting scheme to separately handle the right and left hand sides of the equation (Gallouet, Laborde, Monsaingeon 2019).

# Is it really not a gradient flow?

- Integrating the equation against the pressure gives the energy “dissipation” inequality:

$$\begin{aligned} & \int_{\Omega} e(\rho(t, x)) dx + \int_0^t \int_{\Omega} \rho(s, x) |\nabla p(s, x)|^2 dx ds \\ & \leq \int_{\Omega} e(\rho(0, x)) dx + \int_0^t \int_{\Omega} p(s, x) \rho(s, x) G(p(s, x)) dx ds \end{aligned}$$

- The space-time integral on the second line should not show up in a  $W^2$  gradient flow.
- Energy is not necessarily being dissipated! We have to pay for adding mass to the system.



Given a time step  $\tau > 0$  we iterate

$$\rho^{n+1} = \operatorname{argmin}_{\rho} \min_{\alpha} \int_{\Omega} e(\rho) + \tau \int_{\Omega} \rho^n f(\alpha) + \frac{1}{2\tau} W_2^2(\rho, \rho^n(1 + \tau\alpha)) \quad (3)$$

- No issue that  $\rho^{n+1}$  and  $\rho^n$  have different mass since we are matching  $\rho^{n+1}$  to  $\rho^n(1 + \tau\alpha^{n+1})$ .
- $f$  is a convex term (linked to  $G$ ) encouraging mass change (can be taken to be spatially dependent).

# Duality and $c$ -transforms

- All of our analysis and numerics will be based on the dual rather than the primal problem.
- We will constantly use the following notation/results:

$c$ -transform:

$$p^c(y) := \inf_x p(x) + \frac{|y - x|^2}{2\tau}.$$

conjugate  $c$ -transform:

$$q^{\bar{c}}(x) = \sup_y q(y) - \frac{|y - x|^2}{2\tau}.$$

Induced transport maps:

$$T_p(y) := \operatorname{argmin}_x p(x) + \frac{|y - x|^2}{2\tau}, \quad \bar{T}_q(x) := \operatorname{argmax}_y q(y) - \frac{|y - x|^2}{2\tau}.$$

$c$ -transform variation:

$$\lim_{t \rightarrow 0^+} \frac{(p + tu)^c(y) - p^c(y)}{\tau} = u(T_p(y)).$$

# Minimax formulation

Using Kantorovich duality

$$\frac{1}{2\tau} W_2^2(\rho, \rho^n(1 + \tau\alpha)) = \int_{\Omega} p^c \rho^n(1 + \tau\alpha) - p\rho$$

We transform the original primal problem

$$\min_{\alpha, \rho} \int_{\Omega} e(\rho) + \tau \rho^n f(\alpha) + \frac{1}{2\tau} W_2^2(\rho, \rho^n(1 + \tau\alpha))$$

into the minimax problem

$$\min_{\rho, \alpha} \sup_p \int_{\Omega} p^c \rho^n(1 + \tau\alpha) + e(\rho) - p\rho + \tau \rho^n f(\alpha)$$

Fixing  $p$  and minimizing over  $\rho$  we get

$$\min_{\alpha} \sup_p \int_{\Omega} p^c \rho^n(1 + \tau\alpha) - e^*(p) + \tau \rho^n f(\alpha)$$

# Determining the growth rate

If we fix  $p$  and minimize over  $\alpha$  in the minimax problem

$$\min_{\alpha} \sup_p \int_{\Omega} p^c \rho^n (1 + \tau \alpha) - e^*(p) + \tau \rho^n f(\alpha)$$

the optimality condition for  $\alpha$  requires

$$f'(\alpha) + p^c = 0 \quad \implies \quad \alpha = f^{*'}(-p^c).$$

- Choosing  $f$  such that  $f^{*'}(-a) = G(a)$  allows us to obtain the desired growth rate.
- As long as  $G$  is decreasing there exists a convex function  $f$  with this property.

# The dual problem

Plugging in the optimal choice for  $\alpha$ , we obtain the maximization problem

$$\sup_p \int_{\Omega} \rho^n(p^c + \tau \bar{G}(p^c)) - e^*(p) \quad (4)$$

where  $\bar{G}$  is an antiderivative of  $G$ .

In the case  $e(\rho) = \frac{1}{\gamma+1} \rho^{\gamma+1}$  we get the dual problem

$$\sup_p \int_{\Omega} \rho^n(p^c + \tau \bar{G}(p^c)) - \frac{\gamma}{\gamma+1} \max(p^{1+\frac{1}{\gamma}}, 0)$$

# The dual problem

- For any function  $p$  we have  $p^{c\bar{c}} \leq p$  and  $p^c = p^{c\bar{c}c}$ .
- $e^*$  is always an increasing function therefore

$$\begin{aligned} & \sup_p \int_{\Omega} \rho^n(p^c + \tau \bar{G}(p^c)) - e^*(p) \\ & \leq \sup_p \int_{\Omega} \rho^n(p^{c\bar{c}c} + \tau \bar{G}(p^{c\bar{c}c})) - e^*(p^{c\bar{c}}) \end{aligned}$$

- Any maximizer  $\bar{p}$  of the dual problem must be a  $c$ -concave function i.e.  $\bar{p}^{c\bar{c}} = \bar{p}$ .
- If we let  $q = p^c$  then we can write the dual problem in the following equivalent way

$$\sup_q \int_{\Omega} \rho^n(q + \tau \bar{G}(q)) - e^*(q^{\bar{c}})$$

# Optimality for the dual problem

$p_{n+1}$  is a solution to the dual problem if

$$T_{p_{n+1} \#}(\rho^n(1 + \tau G(p_{n+1}^c))) = e^{*'}(p_{n+1})$$

If we choose

$$\bar{\rho} = T_{p_{n+1} \#}(\rho^n(1 + \tau G(p_{n+1}^c))), \quad \bar{\alpha} = G(-p_{n+1}^c)$$

then

$$J(p_{n+1}) = F(\bar{\rho}, \bar{\alpha})$$

where  $F$  and  $J$  denote the values of the primal and dual problems respectively.

# Relations and the approximate equation

Combining our work the primal and dual variables satisfy the equations

$$\begin{cases} T_{p_{n+1}} \# (\rho^n (1 + \tau G(p_{n+1}^c))) = \rho^{n+1}, \\ \rho^{n+1} = e^{*'}(p_{n+1}) \iff p_{n+1} = e'(\rho^{n+1}) \end{cases}$$

Rewriting

$$T_{p_{n+1}}^{-1} \rho^{n+1} = \rho^n (1 + \tau G(p_{n+1}^c))$$

we see that given a smooth test function  $\varphi$

$$\int_{\Omega} \frac{\rho^{n+1} - \rho^n}{\tau} \varphi = \int_{\Omega} \rho^{n+1} \frac{\varphi - \varphi \circ T_{p_{n+1}}^{-1}}{\tau} + \varphi \rho^n G(p_{n+1}^c)$$



# Relations and the approximate equation

Since

$$T_{\rho_{n+1}}^{-1} = x + \tau \nabla p_{n+1},$$

We have

$$\begin{aligned} \int_{\Omega} \rho^{n+1} \frac{\varphi - \varphi \circ T_{\rho_{n+1}}^{-1}}{\tau} &= \int_{\Omega} -\rho^{n+1} \nabla \varphi \cdot \nabla p_{n+1} \\ &+ O(\tau \|D^2 \varphi\|_{L^\infty(\Omega)} \|\nabla p\|_{L^2(\Omega)}^2) \end{aligned}$$

As long as we can control  $\|\nabla p_{n+1}\|_{L^2(\Omega)}^2$  and prove the weak convergence of the nonlinear terms  $e'(\rho^{n+1})$ ,  $\rho^{n+1} \nabla p_{n+1}$  and  $G(p_{n+1}^c)$  then the scheme converges to a weak solution of the tumor growth model.

## Theorem (J., Kim, Tong 2021)

Given two densities  $\rho_0, \rho_1$  let

$$\bar{\rho}_i = \operatorname{argmin}_{\rho} \int_{\Omega} e(\rho) + \tau \rho_i f(\alpha) + \frac{1}{2\tau} W_2^2(\rho, \rho_i(1 + \tau\alpha)).$$

If  $\rho_0 \leq \rho_1$  a.e., then  $\bar{\rho}_0 \leq \bar{\rho}_1$  a.e.

To prove this we work with the dual problem. If we let

$$\bar{\rho}_i = \operatorname{argmax}_p \int_{\Omega} \rho_i(p^c + \tau \bar{G}(p^c)) - e^*(p)$$

then it will be enough to show that  $\bar{\rho}_0 \leq \bar{\rho}_1$  a.e.

Recall that

$$T_p(y) = \operatorname{argmin}_x p(x) + \frac{|y - x|^2}{2\tau}$$

Lemma (J., Kim, Tong 2021)

Let  $p_0, p_1$  be  $c$ -concave functions and let  $U = \{x \in \Omega : p_0(x) > p_1(x)\}$ . If  $T_{p_0}(y) \in U$ , then  $T_{p_1}(y) \in U$  and  $p_1^c(y) \leq p_0^c(y)$ .

Remark: This doesn't use any property of the quadratic cost beyond existence of optimal maps.

# Comparison principle

Let  $\chi$  be the characteristic function of  $U = \{x \in \Omega : \bar{\rho}_0(x) > \bar{\rho}_1(x)\}$ . Optimality of the  $\bar{\rho}_i$  implies that

$$\int_{\Omega} \chi e^{*'}(\bar{\rho}_i) = \int_{\Omega} \rho_i (1 + \tau G(\bar{\rho}_i^c)) \chi \circ T_{\bar{\rho}_i}$$

Therefore we have the chain of inequalities

$$\begin{aligned} \int_{\Omega} \chi e^{*'}(\bar{\rho}_0) &\geq \int_{\Omega} \chi e^{*'}(\bar{\rho}_1) = \int_{\Omega} \rho_1 (1 + \tau G(\bar{\rho}_1^c)) \chi \circ T_{\bar{\rho}_1} \\ &\geq \int_{\Omega} \rho_1 (1 + \tau G(\bar{\rho}_0^c)) \chi \circ T_{\bar{\rho}_0} \geq \int_{\Omega} \rho_0 (1 + \tau G(\bar{\rho}_0^c)) \chi \circ T_{\bar{\rho}_0} = \int_{\Omega} \chi e^{*'}(\bar{\rho}_0) \end{aligned}$$

# Numerics (Back-and-Forth Method J. Léger 2020, J. Léger, Lee 2021)

- Evolve the scheme by solving the dual problem using BFM.
- BFM performs alternating  $H^1$  gradient ascent on the two equivalent dual problems:

$$J_n(p) = \int_{\Omega} \rho^n (p^c + \tau \bar{G}(p^c)) - e^*(p), \quad I_n(q) = \int_{\Omega} \rho^n (q + \tau \bar{G}(q)) - e^*(q^{\bar{c}})$$

- Once we have recovered the optimal pressure  $p_{n+1}$  we can recover the optimal density  $\rho^{n+1}$  through either of the relations

$$\rho^{n+1} = e^{*'}(p_{n+1}), \quad \rho^{n+1} = T_{p_{n+1}} \# \left( \rho^n (1 + \tau G(p_{n+1}^c)) \right).$$

Given an initial pressure  $p^0$  and step size  $\sigma$  we iterate

$$p^{k+1/2} = p^k + \sigma \nabla_{H_1} J_n(p^k),$$

$$q^{k+1/2} = (p^{k+1/2})^c,$$

$$q^{k+1} = q^{k+1/2} + \sigma \nabla_{H_1} I_n(q^{k+1/2}),$$

$$p^{k+1} = (q^{k+1})^{\bar{c}}.$$

- $H^1$  gradient is equivalent to preconditioning the standard  $L^2$  gradient by  $(I - \Delta)^{-1}$

$$\nabla_{H_1} J_n(p) = (I - \Delta)^{-1} \left( T_{p\#} \rho^n (1 + \tau G(p^c)) - e^{*'}(p) \right)$$

$$\nabla_{H_1} I_n(q) = (I - \Delta)^{-1} \left( \rho^n (1 + \tau G(q)) - \bar{T}_{q\#} e^{*'}(q^{\bar{c}}) \right)$$

- $H^1$  is the weakest inner product where  $I$  and  $J$  have a hope of being  $L$ -smooth for some  $L < \infty$  ( $c$ -transform is not stable in weaker norms).
- Functions like  $\max(p, 0)$  that are not  $L^2$  smooth are  $H^1$  smooth as long as  $\partial\{p > 0\}$  is nondegenerate (trace inequality!)
- Alternating between  $I$  and  $J$  is beneficial because their Hessians are almost inverses of one another.
- On a grid with  $N$  points the  $c/\bar{c}$ -transforms as well as the induced maps  $T_p$  and  $\bar{T}_q$  can be computed in  $O(N)$  time.

# Simulations