

Generalized Bregman Distances

Regina S. Burachik

University of South Australia

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Outline

- 1 Classical Bregman distance
 - Definition
 - Examples
- 2 Generalized Bregman distances
 - Definition
 - The Fitzpatrick case
 - Examples
 - Generalized Bregman envelopes
 - Definition for two maps
 - Applications
- 3 Open problems and Conclusion



This talk contains joint work with:

[Juan Enrique Martínez Legaz](#)

Universitat Autònoma de Barcelona, Spain

[Minh N. Dao](#)

Federation University, Ballarat, Australia

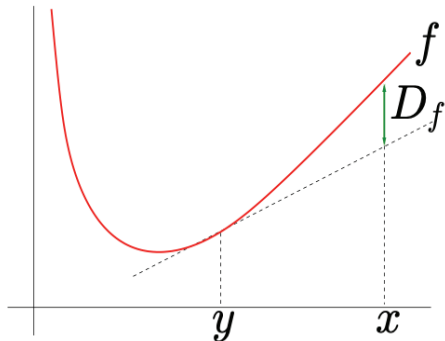
[Scott B. Lindstrom](#)

Hong Kong Polytechnic and Curtin University, Perth, Australia



$f : X \rightarrow \mathbb{R}_{+\infty}$ strictly convex, smooth, (Bregman, 1967)

Bregman distance : $D_f(x, y) := f(x) - f(y) - \langle \nabla f(y), x - y \rangle$



Using specific choices of f , we obtain:

Kullback-Liebler divergence A distance between positive vectors, used in information theory, statistics, portfolio selection,...

Itakura-Saito divergence A non-symmetric measure of difference between probability distributions. Used to measure sound quality and speech processing

Squared Euclidean (ℓ_2) distance for $f(\cdot) := (1/2)\|\cdot\|^2$ we have
$$D_f(x, y) = (1/2)\|x - y\|^2$$



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What is the Bregman distance really measuring?

$$D_f(x, y) = f(x) - f(y) - \langle \nabla f(y), x - y \rangle =$$

$$\text{(Fenchel - Young)} = f(x) + f^*(\nabla f(y)) - \langle \nabla f(y), x \rangle$$

Hence, $D_f(x, y) = 0 \iff \nabla f(x) = \nabla f(y)$

so D_f measures distance between images of the gradient map!



If f is smooth, then

$$D_f(x, y) \geq 0 \quad \forall x, y \iff f \text{ is convex}$$

Due to a “Fenchel-Young” property for the graph of ∇f :

$$f(x) + f^*(v) \geq \langle x, v \rangle \quad \forall x \in X, v \in X^*$$

$$f(x) + f^*(v) = \langle x, v \rangle \iff v = \nabla f(x)$$

An analogous property holds for max-mon T !



The family $\mathcal{H}(T)$ Introduced in [RSB-Svaiter, 2002]

For $T : X \rightrightarrows X^*$ max-mon, the **Fitzpatrick family** $\mathcal{H}(T)$ consists of functions $h : X \times X^* \rightarrow \mathbb{R}_{+\infty}$ convex and norm-weak* lsc, s.t.

$$\begin{aligned} h(x, v) &\geq \langle x, v \rangle \quad \forall x \in X, v \in X^* \\ h(x, v) &= \langle x, v \rangle \iff v \in Tx, \end{aligned}$$

Fenchel-Young gives $h(x, v) := f(x) + f^*(v) \in \mathcal{H}(\partial f)$



$\mathcal{H}(T)$ and the Fitzpatrick function

$\mathcal{H}(T)$ has a smallest and a biggest element. The smallest one is the *Fitzpatrick function*¹:

$$\mathcal{F}_T(x, v) := \sup_{(z, w) \in G(T)} \langle z - x, v - w \rangle + \langle x, v \rangle.$$

The biggest is $\sigma_T = \text{cl conv}(\pi + \delta_{G(T)})$ and in fact

$$\sigma_T(x, v) := \mathcal{F}_T^*(v, x),$$

where $\pi := \langle \cdot, \cdot \rangle$.

¹Fitzpatrick, 1988.

Extending the Bregman distance to a generic T

For T max-mon, $x, y \in D(T)$, and $h \in \mathcal{H}(T)$, recall that

$$\begin{aligned} h(x, v) &\geq \langle x, v \rangle \quad \forall x \in X, v \in X^* \\ h(x, v) &= \langle x, v \rangle \iff v \in Tx, \end{aligned}$$

The "sharp" version

$$D_T^{\sharp, h}(x, y) := \sup_{v \in Ty} h(x, v) - \langle x, v \rangle$$

The "flat" version

$$D_T^{\flat, h}(x, y) := \inf_{v \in Ty} h(x, v) - \langle x, v \rangle$$

Note that the distance depends on the choice of h !



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Particular case 1: $T = \nabla f$ and $h = h^{FY} = f + f^*$

If f convex and smooth, choose then

$$D_{\nabla f}^{\#, h^{FY}}(x, y) = D_{\nabla f}^{b, h^{FY}}(x, y) = D_f(x, y),$$

for $(x, y) \notin (\text{dom } f \setminus \text{dom } \nabla f) \times \text{dom } \nabla f$ (region where coincides with the classical Bregman distance)



Particular case 2: T max-mon and $h = \mathcal{F}_T$

The “sharp” version

$$D_T^{\sharp, h}(x, y) := \sup_{v \in T y} \mathcal{F}_T(x, v) - \langle x, v \rangle$$

The “flat” version

$$D_T^b(x, y) := \inf_{v \in T y} \mathcal{F}_T(x, v) - \langle x, v \rangle$$

where

$$\mathcal{F}_T(x, v) := \sup_{(z, w) \in G(T)} \langle z - x, v - w \rangle + \langle x, v \rangle.$$

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$$T = \nabla f \text{ and } h = \mathcal{F}_{\nabla f}$$

Example 1. Negative Burg entropy

$f : \mathbb{R} \rightarrow \mathbb{R}$ defined as

$$f(t) = \begin{cases} -\log t, & t > 0 \\ +\infty, & \text{c.c.} \end{cases}$$

$$D_{\nabla f}^{\mathcal{F}_{\nabla f}}(x, y) = \begin{cases} \left(\sqrt{\frac{x}{y}} - 1 \right)^2, & x \geq 0, y > 0 \\ +\infty, & \text{c.c.} \end{cases}$$

Used $\mathcal{F}_{\nabla f}$ from [Bauschke-McLaren-Sendov, 2005]



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Example 2: negative Boltzmann-Gibbs-Shannon/Kullback-Leibler/ entropy

$$f : \mathbb{R} \rightarrow \mathbb{R} \text{ defined as } f(t) = \begin{cases} t \log t - t, & t > 0 \\ 0 & t = 0 \\ +\infty, & \text{c.c.} \end{cases}$$

$$D_{\nabla f}^{\mathcal{F}_{\nabla f}}(x, y) = \begin{cases} \frac{y}{e}, & x = 0, y \geq 0 \\ x \left[W\left(e \frac{x}{y}\right) + \frac{1}{W\left(e \frac{x}{y}\right)} - 2 \right], & x > 0, y > 0 \\ +\infty, & \text{c.c.} \end{cases}$$

$W : \mathbb{R}_{++} \rightarrow \mathbb{R}_{++}$ inverse of $t \rightarrow te^t$ (Lambert-W function). Used $\mathcal{F}_{\nabla f}$ from [Bauschke-McLaren-Sendov, 2005]



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Prox-envelopes: Moreau, Bregman, Fitzpatrick

The **Moreau** prox-envelope of a convex $\theta : \mathbb{R} \rightarrow \mathbb{R}$:

$$e_{\gamma}^M(\theta)(x) := \inf_w \left\{ \theta(w) + \frac{1}{\gamma} \|w - x\|^2 \right\}$$

The **Bregman** prox-envelope:

$$e_{\gamma}^B(\theta)(x) := \inf_w \left\{ \theta(w) + \frac{1}{\gamma} D_f(w, x) \right\},$$

where $D_f(x, y) = D_{\nabla f}^{h^{FY}}(x, y) = D_{\nabla f}^{(f+f^*)}(x, y)$ classical Bregman distance



A new family of Prox-envelopes

The **Fitzpatrick** prox-envelope:

$$e_{\gamma}^F(\theta)(x) := \inf_w \left\{ \theta(w) + \frac{1}{\gamma} D_{\nabla f}^{\mathcal{F}\nabla f}(w, x) \right\}$$

or, in general, taking $D_T^{*,h} \in \{D_T^{\sharp,h}, D_T^{b,h}\}$:

$$e_{\gamma}^{*,h,T}(\theta)(x) := \inf_w \left\{ \theta(w) + \frac{1}{\gamma} D_T^{*,h}(w, x) \right\}$$



Asymptotic behaviour when $\gamma \downarrow 0$:

- θ convex, proper, lsc, $y \in \text{dom } T \cap \text{dom } \theta$
- $(\text{dom } T \cap \text{dom } \theta) \times \{y\} \subseteq \text{dom } \mathcal{D}_T^{\sharp, h}$
- $\varphi_\mu(\cdot) := \theta(\cdot) + \frac{1}{\mu} \mathcal{D}_T^{\sharp, h}(\cdot, y)$ is coercive for some $\mu \in \mathbb{R}_{++}$.

Let $s_\gamma \in \text{Argmin } \varphi_\gamma$. Then, as $\gamma \downarrow 0$,

- $\mathcal{D}_T^{\sharp, h}(s_\gamma, y) \rightarrow 0$.
- If T is str. mon. over $\text{dom } T \cap \text{dom } \theta$,

$$s_\gamma \rightarrow y, \quad e_{\frac{\sharp, h, T}{\gamma}}(\theta)(y) \uparrow \theta(y), \quad \theta(s_\gamma) \rightarrow \theta(y), \quad \frac{1}{\gamma} \mathcal{D}_T^{\sharp, h}(s_\gamma, y) \rightarrow 0.$$



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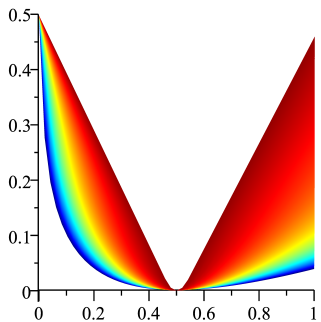
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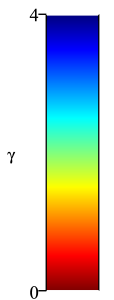


Bregman vs. Fitzpatrick envelope for $\theta(t) := |t - (1/2)|$ for Kullback-Leibler f , $T = \log(\cdot)$

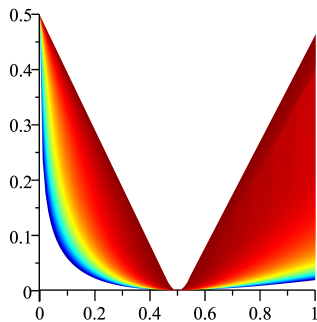


Bregman envelope

$$h = f + f^*$$



color key



Fitzpatrick envelope

$$h = \mathcal{F}_{\nabla f}$$

*Credit for the figures: Scott Lindstrom



Bregman distance between two set-valued maps

$T, S : X \rightrightarrows X^*$, T max-mon, S gral., $h_T \in \mathcal{H}(T)$:

The “sharp” version

$$D_S^{\sharp, h_T}(x, y) := \sup_{v \in S y} h_T(x, v) - \langle x, v \rangle$$

The “flat” version

$$D_S^b, h_T(x, y) := \inf_{v \in S y} h_T(x, v) - \langle x, v \rangle$$



A new measure of degree of overlap between T_x and S_y

T max-mon, $h \in \mathcal{H}(T)$, $(x, y) \in D(T) \times D(S)$:

(a) If S loc. bded in $\text{int}D(S)$, weakly-closed valued,

$(x, y) \notin \text{bdry}D(S) \times \text{bdry}D(T)$.

$$D_S^{b,h}(x, y) = 0 \iff S_y \cap T_x \neq \emptyset$$

$$(b) \quad D_S^{\sharp,h}(x, y) = 0 \iff S_y \subset T_x$$



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T max-mon, $h \in \mathcal{H}(T)$, $u \in X^*$: Find x s.t. $u \in Tx$ (P_T)

If (P_T) difficult, find y sol. of the “better conditioned” problem

$$(P_S) \quad u \in Sy$$

For this y , find x s.t. $D_S^{\sharp, h}(x, y) \leq \varepsilon$, then x solves

$$(P_{T, \varepsilon}) \quad u \in T^e(\varepsilon, x),$$

where

$$T^e(\varepsilon, x) := \{v \in X^* : \langle x - y, v - w \rangle \geq -\varepsilon \forall (y, w) \in G(T)\}$$

Note: $D_S^{\sharp, h}(\cdot, y)$ is convex and x solves an ε -approximation of (P_T)!



An enlargement of T induced by $h \in \mathcal{H}(T)$

T max-mon, the *enlargement* L^h of T is

$$L^h(\varepsilon, x) := \{v \in X^* : h(x, v) - (x, v) \leq \varepsilon\}.$$

Consider the problem:

$$(PS) \text{ find } x \in X \text{ s.t. } 0 \in Sx + Tx.$$

GBDs define approximate solutions of (PS) using L^h



T max-mon, S point-to-set, $h \in \mathcal{H}(T)$

Consider the following statements:

(a) $0 \in L^h(\varepsilon, x) + Sx.$

(b) $\mathcal{D}_{-S}^{b,h}(x, x) \leq \varepsilon.$

Then (a) \implies (b) (necessary condition for optimality). Moreover, if $\text{dom } S$ is open and S is locally bounded with weakly closed images, then the two statements are equivalent.



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Some Questions

- (a) How can we use these new distances for obtaining more efficient solution techniques for variational inequalities/inclusion problems
- (b) When $T = S = \partial f$, will the Fitzpatrick distances play a role similar to that of the classical Bregman distances (Bregman projections, convergence analysis, etc)?
- (c) Can these distances be used to regularize/penalize in prox-like iterations for variational inequalities?



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Conclusions

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Some notions involving convex functions (such as classical Bregman distances), can be extended to maximally monotone operators, thus producing new tools both for convex analysis and for maximal monotone theory.






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