# Entropic Projections and Dominating Points 

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# ENTROPIC PROJECTIONS AND DOMINATING POINTS 

CHRISTIAN LÉONARD


#### Abstract

Entropic projections and dominating points are solutions to convex minimization problems related to conditional laws of large numbers. They appear in many areas of applied mathematics such as statistical physics, information theory, mathematical statistics, ill-posed inverse problems or large deviation theory. By means of convex conjugate duality and functional analysis, criteria are derived for the existence of entropic projections, generalized entropic projections and dominating points. Representations of the generalized entropic projections are presented. It is shown that they are the "measure component" of some extended entropy minimization problem. This approach leads to new results and offers a new point of view. It also permits to extend previous results on the subject by removing unnecessary topological restrictions. As a by-product, new proofs of already known results are provided.


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## 1. Introduction

Entropic projections and dominating points are solutions to convex minimization problems related to conditional laws of large numbers. They appear in many areas of applied mathematics such as statistical physics, information theory, mathematical statistics, illposed inverse problems or large deviation theory.

Conditional laws of large numbers. Suppose that the empirical measures

$$
\begin{equation*}
L_{n}:=\frac{1}{n} \sum_{i=1}^{n} \delta_{Z_{i}}, \quad n \geq 1, \tag{1.1}
\end{equation*}
$$

of the $\mathcal{Z}$-valued random variables $Z_{1}, Z_{2}, \ldots\left(\delta_{z}\right.$ is the Dirac measure at $\left.z\right)$ obey a Large Deviation Principle (LDP) in the set $P_{\mathcal{Z}}$ of all probability measures on $\mathcal{Z}$ with the rate

[^0]function $I$. This approximately means that $\mathbb{P}\left(L_{n} \in \mathcal{A}\right) \underset{n \rightarrow \infty}{\asymp} \exp \left[-n \inf _{P \in \mathcal{A}} I(P)\right]$ for $\mathcal{A} \subset P_{\mathcal{Z}}$. With regular enough subsets $\mathcal{A}$ and $\mathcal{C}$ of $P_{\mathcal{Z}}$, one can expect that for "all" $\mathcal{A}$
\[

\lim _{n \rightarrow \infty} \mathbb{P}\left(L_{n} \in \mathcal{A} \mid L_{n} \in \mathcal{C}\right)= $$
\begin{cases}1, & \text { if } \mathcal{A} \ni P_{*} \\ 0, & \text { otherwise }\end{cases}
$$
\]

where $P_{*}$ is a minimizer of $I$ on $\mathcal{C}$. To see this, remark that (formally) $\mathbb{P}\left(L_{n} \in \mathcal{A} \mid L_{n} \in\right.$ $\mathcal{C}) \underset{n \rightarrow \infty}{\asymp} \exp \left[-n\left(\inf _{P \in \mathcal{A} \cap \mathcal{C}} I(P)-\inf _{P \in \mathcal{C}} I(P)\right)\right]$. If $I$ is strictly convex and $\mathcal{C}$ is convex, $P_{*}$ is unique and this roughly means that conditionally on $L_{n} \in \mathcal{C}$, as $n$ tends to infinity $L_{n}$ tends to the solution $P_{*}$ of the minimization problem

$$
\begin{equation*}
\text { minimize } I(P) \text { subject to } P \in \mathcal{C}, \quad P \in P_{\mathcal{Z}} \tag{1.2}
\end{equation*}
$$

Such conditional Laws of Large Numbers (LLN) appear in information theory and in statistical physics where they are often called Gibbs conditioning principles (see $\sqrt{7}$, Section 7.3] and the references therein). If the variables $Z_{i}$ are independent and identically distributed with law $R$, the LDP for the empirical measures is given by Sanov's theorem and the rate function $I$ is the relative entropy

$$
I(P)=I(P \mid R)=\int_{\mathcal{Z}} \log \left(\frac{d P}{d R}\right) d P, \quad P \in P_{\mathcal{Z}}
$$

Instead of the empirical probability measure of a random sample, one can consider another kind of random measure. Let $z_{1}, z_{2}, \ldots$ be deterministic points in $\mathcal{Z}$ such that the empirical measure $\frac{1}{n} \sum_{i=1}^{n} \delta_{z_{i}}$ converges to $R \in P_{\mathcal{Z}}$. Let $W_{1}, W_{2}, \ldots$ be a sequence of independent random real variables. The random measure of interest is

$$
\begin{equation*}
L_{n}=\frac{1}{n} \sum_{i=1}^{n} W_{i} \delta_{z_{i}} \tag{1.3}
\end{equation*}
$$

where the $W_{i}$ 's are interpreted as random weights. If the weights are independent copies of $W$, as $n$ tends to infinity, $L_{n}$ tends to the deterministic measure $\mathbb{E} W \cdot R$ and obeys the LDP in the space $M_{\mathcal{Z}}$ of measures on $\mathcal{Z}$ with rate function $I(Q)=\int_{\mathcal{Z}} \gamma^{*}\left(\frac{d Q}{d R}\right) d R, Q \in M_{\mathcal{Z}}$ where $\gamma^{*}$ is the Cramér transform of the law of $W$. In case the $W_{i}$ 's are not identically distributed, but have a law which depends (continuously) on $z_{i}$, one can again show that under additional assumptions $L_{n}$ obeys the LDP in $M_{\mathcal{Z}}$ with rate function

$$
I(Q)=\left\{\begin{array}{ll}
\int_{\mathcal{Z}} \gamma_{z}^{*}\left(\frac{d Q}{d R}(z)\right) R(d z), & \text { if } Q \prec R  \tag{1.4}\\
+\infty, & \text { otherwise }
\end{array}, \quad Q \in M_{\mathcal{Z}}\right.
$$

where $\gamma_{z}^{*}$ is the Cramér transform of $W_{z}$. As $\gamma^{*}$ is the convex conjugate of the log-Laplace transform of $W$, it is a convex function: $I$ is a convex integral functional. It is often called an entropy. Again, conditional LLNs hold for $L_{n}$ and lead to the entropy minimization problem:

$$
\begin{equation*}
\text { minimize } I(Q) \text { subject to } Q \in \mathcal{C}, \quad Q \in M_{\mathcal{Z}} \tag{1.5}
\end{equation*}
$$

The large deviations of these random measures and their conditional LLNs enter the framework of Maximum Entropy in the Mean (MEM) which has been studied among others by Dacunha-Castelle, Csiszár, Gamboa, Gassiat, Najim see [5, 6, 10, 21] and also [7, Theorem 7.2.3]. This problem also arises in the context of statistical physics. It has been studied among others by Boucher, Ellis, Gough, Puli and Turkington, see [1] (9].

The relative entropy corresponds to $\gamma^{*}(t)=t \log t-t+1$ in (1.4): the problem (1.2) is a special case of (1.5) with the additional constraint that $Q(\mathcal{Z})=1$.

In this paper, the constraint set $\mathcal{C}$ is assumed to be convex as is almost always done in the literature on the subject. This allows to rely on convex analysis, saddle-point theory or on the geometric theory of projection on convex sets.

Entropic projections. The minimizers of (1.5) are called entropic projections. It may happen that even if the minimizer is not attained, any minimizing sequence converges to some measure $Q_{*}$ which does not belong to $\mathcal{C}$. This intriguing phenomenon was discovered by Csiszár [3]. Such a $Q_{*}$ is called a generalized entropic projection.

In the special case where $I$ is the relative entropy, Csiszár has obtained existence results in (2) together with dual equalities. His proofs are based on geometric properties of the relative entropy; no convex analysis is needed. Based on the same geometric ideas, he obtained later in [3] a powerful Gibbs conditioning principle for noninteracting particles. For general entropies as in (1.4), he studies the problem of existence of entropic and generalized entropic projections in [⿴囗

The minimization problem (1.5) is interesting in its own right, even when conditional LLNs are not at stake. The literature on this subject is huge. Some bibliographical entries are given in [15].
Dominating points. Let the constraint set $\mathcal{C}$ be described by

$$
\begin{equation*}
\mathcal{C}=\left\{Q \in M_{\mathcal{Z}} ; T Q \in C\right\} \tag{1.6}
\end{equation*}
$$

where $C$ is a subset of a vector space $\mathcal{X}$ and $T: M_{\mathcal{Z}} \rightarrow \mathcal{X}$ is a linear operator. As a typical example, one can think of $T Q=\int_{\mathcal{Z}} \theta(z) Q(d z)$ where $\theta: \mathcal{Z} \rightarrow \mathcal{X}$ is some function and the integral should be taken formally for the moment. With $L_{n}$ given at (1.1) or (1.3), if $T$ is regular enough, we obtain by the contraction principle that $X_{n}:=T L_{n}=\frac{1}{n} \sum_{i=1}^{n} \theta\left(Z_{i}\right) \in \mathcal{X}$ or $X_{n}:=T L_{n}=\frac{1}{n} \sum_{i=1}^{n} W_{i} \theta\left(z_{i}\right)$ obeys the LDP in $\mathcal{X}$ with rate function $J(x)=\inf \left\{I(Q) ; Q \in M_{\mathcal{Z}}, T Q=x\right\}, x \in \mathcal{X}$. Once again, the conditional LLN for $X_{n}$ is of the form: For "all" $A \subset \mathcal{X}$,

$$
\lim _{n \rightarrow \infty} \mathbb{P}\left(X_{n} \in A \mid X_{n} \in C\right)= \begin{cases}1, & \text { if } A \ni x_{*} \\ 0, & \text { otherwise }\end{cases}
$$

where $x_{*}$ is a solution to the minimization problem

$$
\begin{equation*}
\text { minimize } J(x) \text { subject to } x \in C, \quad x \in \mathcal{X} \tag{1.7}
\end{equation*}
$$

The minimizers of (1.7) are called dominating points. This notion was introduced by Ney [22, 23] in the special case where $\left(Z_{i}\right)_{i>1}$ is an iid sequence in $\mathcal{Z}=\mathbb{R}^{d}$ and $\theta$ is the identity, i.e. $X_{n}=\frac{1}{n} \sum_{i=1}^{n} Z_{i}$. Later, Einmahl and Kuelbs [8, [13] have extended this study to a Banach space $\mathcal{Z}$. In this iid case, $J$ is the Cramér transform of the law of $Z_{1}$.

Presentation of the results. One treats the problems of existence of entropic projections and dominating points in a unified way, taking advantage of the mapping $T Q=x$. Although it is simple, this connection does not seem to be used in previous literature. Hence, one mainly concentrates efforts on the entropic projections and then transports the results to the dominating points.

It will be proved at Proposition 5.5 that the entropic projection exists on $\mathcal{C}$ if the supporting hyperplanes of $\mathcal{C}$ are directed by sufficiently integrable functions. In some cases of not enough integrable supporting hyperplanes, the representation of the generalized projection is still available and given at Theorem 5.6. It will appear that the generalized projection is the "measure" part of the minimizer of an extended minimization problem.

For instance, with the relative entropy $I(. \mid R)$, the projection exists in $\mathcal{C}$ if its supporting hyperplanes are directed by functions $u$ such that $\int_{\mathcal{Z}} e^{\alpha|u|} d R<\infty$ for all $\alpha>0$, see

Proposition 5.10. If these $u$ only satisfy $\int_{\mathcal{Z}} e^{\alpha|u|} d R<\infty$ for some $\alpha>0$, the projection may not exist in $\mathcal{C}$, but the generalized projection is computable: its Radon-Nykodym derivative with respect to $R$ is characterized at Proposition 5.13.

One finds again some already known results of Csiszár 3, 4, U. Einmahl and Kuelbs [8, 13] with different proofs and a new point of view. The representations of the generalized projections are new results. The conditions on $C$ to obtain dominating points are improved and an interesting phenomenon noticed in [13] is clarified at Remark 6.9 by connecting it with the generalized entropic projection.

The main results are Theorems 4.1, 4.7, 5.6 and 6.8.

Outline of the paper. At Section 2, one gives a precise formulation of the entropy minimization problem (1.5) and a natural extension of it, see $\left(\overline{\mathbb{P}_{C}}\right)$ at Section 2.3. Then one recalls at Theorems 2.14 and 2.17 results from [15] about the existence and uniqueness of the solutions of (1.5) and $(\mathbb{P} d)$, related dual equalities and the characterizations of their solutions in terms of integral representations.
Examples of standard entropies and constraints are presented at Section 3 .
One shows at Theorem 4.7 in Section 4 that under "bad" constraints (this notion is specified at Section 2.4 in terms of the set $C$ and the operator $T$ appearing in (1.6)), although the problem (1.5) may not be attained, its minimizing sequences may converge in some sense to some measure $Q_{*}$ : the generalized entropic projection.

Section 5 is mainly a restatement of Sections 2 and 1 in terms of entropic projections. The results are also stated explicitly for the special important case of the relative entropy.

Section 6 is devoted to dominating points. As they are continuous images of entropic projections, the main results of this section are corollaries of the results of Section 5 .

Notation. Let $X$ and $Y$ be topological vector spaces. The algebraic dual space of $X$ is $X^{*}$, the topological dual space of $X$ is $X^{\prime}$. The topology of $X$ weakened by $Y$ is $\sigma(X, Y)$ and one writes $\langle X, Y\rangle$ to specify that $X$ and $Y$ are in separating duality.
Let $f: X \rightarrow[-\infty,+\infty]$ be an extended numerical function. Its convex conjugate with respect to $\langle X, Y\rangle$ is $f^{*}(y)=\sup _{x \in X}\{\langle x, y\rangle-f(x)\} \in[-\infty,+\infty], y \in Y$. Its subdifferential at $x$ with respect to $\langle X, Y\rangle$ is $\partial_{Y} f(x)=\{y \in Y ; f(x+\xi) \geq f(x)+\langle y, \xi\rangle, \forall \xi \in X\}$. If no confusion occurs, one writes $\partial f(x)$.
Let $A$ be a subset of $X$, its intrinsic core is icor $A=\left\{x \in A ; \forall x^{\prime} \in \operatorname{aff} A, \exists t>0,[x, x+\right.$ $t\left(x^{\prime}-x\right)[\subset A\}$ where aff $A$ is the affine space spanned by $A$. Let us denote $\operatorname{dom} f=\{x \in$ $X ; f(x)<\infty\}$ the effective domain of $f$ and icordom $f$ the intrisic core of dom $f$. The indicator of a subset $A$ of $X$ is defined by

$$
\iota_{A}(x)=\left\{\begin{array}{ll}
0, & \text { if } x \in A \\
+\infty, & \text { otherwise }
\end{array}, \quad x \in X .\right.
$$

One writes

$$
I_{\varphi}(u):=\int_{\mathcal{Z}} \varphi(z, u(z)) R(d z)=\int_{\mathcal{Z}} \varphi(u) d R
$$

and $I=I_{\gamma^{*}}$ for short, instead of (1.4).

## 2. Minimizing entropy under convex constraints

In this section, the main results of (15] are recalled.
2.1. Orlicz spaces. The fact that the generalized projection may not belong to $\mathcal{C}$ is connected with some properties of Orlicz spaces associated to $I$. Let us recall some basic definitions and results.

A set $\mathcal{Z}$ is furnished with a $\sigma$-finite nonnegative measure $R$ on a $\sigma$-field which is assumed to be $R$-complete. A function $\rho: \mathcal{Z} \times \mathbb{R}$ is said to be a Young function if for $R$-almost every $z, \rho(z, \cdot)$ is a convex even $[0, \infty]$-valued function on $\mathbb{R}$ such that $\rho(z, 0)=0$ and there exists a measurable function $z \mapsto s_{z}>0$ such that $0<\rho\left(z, s_{z}\right)<\infty$.
In the sequel, every numerical function on $\mathcal{Z}$ is supposed to be measurable.
Definitions 2.1 (The Orlicz spaces $L_{\rho}$ and $E_{\rho}$ ). The Orlicz space associated with $\rho$ is defined by $L_{\rho}=\left\{u: \mathcal{Z} \rightarrow \mathbb{R} ;\|u\|_{\rho}<+\infty\right\}$ where the Luxemburg norm $\|\cdot\|_{\rho}$ is defined by $\|u\|_{\rho}=\inf \left\{\beta>0 ; \int_{\mathcal{Z}} \rho(z, u(z) / \beta) R(d z) \leq 1\right\}$ and $R$-a.e. equal functions are identified. Hence,

$$
L_{\rho}=\left\{u: \mathcal{Z} \rightarrow \mathbb{R} ; \exists \alpha_{o}>0, \int_{\mathcal{Z}} \rho\left(z, \alpha_{o} u(z)\right) R(d z)<\infty\right\} .
$$

A subspace of interest is

$$
E_{\rho}=\left\{u: \mathcal{Z} \rightarrow \mathbb{R} ; \forall \alpha>0, \int_{\mathcal{Z}} \rho(z, \alpha u(z)) R(d z)<\infty\right\}
$$

Of course $E_{\rho} \subset L_{\rho}$. Note that if $\rho$ doesn't depend on $z$ and $\rho\left(s_{o}\right)=\infty$ for some $s_{o}>0$, $E_{\rho}$ reduces to the null space and if in addition $R$ is bounded, $L_{\rho}$ is $L_{\infty}$. On the other hand, if $\rho$ is a finite function which doesn't depend on $z$ and $R$ is bounded, $E_{\rho}$ contains all the bounded functions.

Duality in Orlicz spaces is intimately linked with the convex conjugacy. The convex conjugate $\rho^{*}$ of $\rho$ is defined by $\rho^{*}(z, t)=\sup _{s \in \mathbb{R}}\{s t-\rho(z, s)\}$. It is also a Young function so that one may consider the Orlicz space $L_{\rho^{*}}$.

A continuous linear form $\ell \in L_{\rho}^{\prime}$ is said to be singular if for all $u \in L_{\rho}$, there exists a decreasing sequence of measurable sets $\left(A_{n}\right)$ such that $R\left(\cap_{n} A_{n}\right)=0$ and for all $n \geq 1$, $\left\langle\ell, u \mathbf{1}_{\mathcal{Z} \backslash A_{n}}\right\rangle=0$. Let us denote $L_{\rho}^{s}$ the subspace of $L_{\rho}^{\prime}$ of all singular forms.
Theorem 2.2 (Representation of $E_{\rho}^{\prime}$ and $L_{\rho}^{\prime}$ ).
(a) Suppose that $\rho$ is a finite Young function. Then,
(i) $E_{\rho}^{\prime}$ is isomorphic to $L_{\rho^{*}}$;
(ii) $\ell \in L_{\rho}^{\prime}$ is singular if and only if $\langle\ell, u\rangle=0$, for all $u$ in $E_{\rho}$.
(b) Let $\rho$ be any Young function. Any $\ell \in L_{\rho}^{\prime}$ is uniquely decomposed as

$$
\begin{equation*}
\ell=\ell^{a}+\ell^{s} \tag{2.3}
\end{equation*}
$$

with $\ell^{a} \in L_{\rho^{*}} R$ and $\ell^{s} \in L_{\rho}^{s}$. This means that $L_{\rho}^{\prime}$ is the direct sum $L_{\rho}^{\prime}=L_{\rho^{*}} R \oplus L_{\rho}^{s}$.
We denote the space of $R$-absolutely continuous signed measures having a density in the Orlicz space $L_{\rho^{*}}$ by $L_{\rho^{*}} R$.

Proof. For a proof of (a), see [11, Thm 4.8] and [12, Proposition 2.1].
About (b): When $L_{\rho}=L_{\infty}$ this result is the usual representation of $L_{\infty}^{\prime}$. When $\rho$ is a finite function, this result is (12], Theorem 2.2). The general result is proved in 24], with $\rho$ not depending on $z$ but the extension to a $z$-dependent $\rho$ is obvious.

In the decomposition (2.3), $\ell^{a}$ is called the absolutely continuous part of $\ell$ while $\ell^{s}$ is its singular part. $\ell^{a}$ is a measure which is absolutely continuous with respect to $R$ and $\ell^{s}$ is not a measure: it is additive but not $\sigma$-additive.

Definition $2.4\left(\Delta_{2}\right.$-condition). The function $\rho$ is said to satisfy the $\Delta_{2}$-condition if there exist $C>0, s_{o} \geq 0$ such that for all $s \geq s_{o}, \rho(2 s) \leq C \rho(s)$.

When $R$ is bounded, in order that $E_{\rho}=L_{\rho}$, it is enough that $\rho$ satisfies the $\Delta_{2}$-condition. If $R$ is unbounded, for $E_{\rho}=L_{\rho}$, it is enough that the $\Delta_{2}$-condition holds globally, i.e. with $s_{o}=0$. Consequently, if $\rho$ satisfies the $\Delta_{2}$-condition we have $L_{\rho}^{\prime}=L_{\rho^{*}} R$ so that $L_{\rho}^{s}$ reduces to the null vector space.

### 2.2. The entropy minimization problem $\left(\mathrm{P}_{C}\right)$.

Entropy. Let $R$ be a positive measure on a space $\mathcal{Z}$ and take a $[0, \infty]$-valued measurable function $\gamma^{*}$ on $\mathcal{Z} \times \mathbb{R}$ such that $\gamma^{*}(z, \cdot):=\gamma_{z}^{*}$ is convex and lower semicontinuous for all $z \in \mathcal{Z}$. Denote $M_{\mathcal{Z}}$ the space of all signed measures $Q$ on $\mathcal{Z}$. The entropy functional to be considered is defined by

$$
I(Q)=\left\{\begin{array}{ll}
\int_{\mathcal{Z}} \gamma_{z}^{*}\left(\frac{d Q}{d R}(z)\right) R(d z) & \text { if } Q \prec R  \tag{2.5}\\
+\infty & \text { otherwise }
\end{array}, \quad Q \in M_{\mathcal{Z}}\right.
$$

where $Q \prec R$ means that $Q$ is absolutely continuous with respect to $R$. Assume that for each $z$ there exists a unique $m(z)$ which minimizes $\gamma_{z}^{*}$ with

$$
\gamma_{z}^{*}(m(z))=0, \forall z \in \mathcal{Z}
$$

Then, $I$ is $[0, \infty]$-valued, its unique minimizer is $m R$ and $I(m R)=0$.
Relevant Orlicz spaces. Since $\gamma_{z}^{*}$ is closed convex for each $z$, it is the convex conjugate of some closed convex function $\gamma_{z}$. Defining

$$
\lambda(z, s)=\gamma(z, s)-m(z) s, \quad z \in \mathcal{Z}, s \in \mathbb{R}
$$

one sees that for $R$-a.e. $z, \lambda_{z}$ is a nonnegative convex function and it vanishes at 0 . Hence,

$$
\lambda_{\diamond}(z, s)=\max [\lambda(z, s), \lambda(z,-s)] \in[0, \infty], \quad z \in \mathcal{Z}, s \in \mathbb{R}
$$

is a Young function and one can consider the corresponding Orlicz spaces $L_{\lambda_{\diamond}}$ and $L_{\lambda_{\circ}^{*}}$ where $\lambda_{\diamond}^{*}(z, \cdot)$ is the convex conjugate of $\lambda_{\diamond}(z, \cdot)$.
The effective domain of $I$ is included in $m R+L_{\lambda_{8}^{*}} R$. It will be assumed from now on that $m \in L_{\lambda_{\circ}^{*}}$ so that $\operatorname{dom} I=L_{\lambda_{\circ}^{*}} R$.
Constraint. In order to define the constraint, take $\mathcal{X}_{o}$ a vector space and a function $\theta: \mathcal{Z} \rightarrow \mathcal{X}_{o}$. One wants to give some meaning to the formal constraint $\int_{\mathcal{Z}} \theta d \ell=x$ with $\ell \in L_{\lambda_{0}^{*}} R$ and $x \in \mathcal{X}_{o}$. Suppose that $\mathcal{X}_{o}$ is the algebraic dual space of some vector space $\mathcal{Y}_{o}$ and define for all $y \in \mathcal{Y}_{o}$ and $z \in \mathcal{Z},\langle y, \theta(z)\rangle_{\mathcal{Y}_{o}, \mathcal{X}_{o}}$, Assuming that

$$
\begin{equation*}
\left\langle\mathcal{Y}_{o}, \theta(\cdot)\right\rangle \subset L_{\lambda_{\diamond}} \tag{2.6}
\end{equation*}
$$

Hölder's inequality in Orlicz spaces allows to define the constraint operator $\ell \in L_{\lambda_{\diamond}}^{\prime} \mapsto$ $\langle\theta, \ell\rangle \in \mathcal{X}_{o}$ by: $\langle y,\langle\theta, \ell\rangle\rangle_{\mathcal{Y}_{o}, \mathcal{X}_{o}}=\langle\langle y, \theta\rangle, \ell\rangle_{L_{\lambda_{\odot}}, L_{\lambda_{o}}^{\prime}}, \forall y \in \mathcal{Y}_{o}$. If $\ell \in L_{\lambda_{\odot}^{*}} R$, one sometimes write $\langle\theta, \ell\rangle=\int_{\mathcal{Z}} \theta d \ell$. One sometimes denote

$$
T \ell=\langle\theta, \ell\rangle, \quad \ell \in L_{\lambda_{\diamond}}^{\prime}
$$

Minimization problem. We are now ready to state the minimization problem (1.5) precisely:

$$
\begin{equation*}
\text { minimize } I(Q) \text { subject to } \int_{\mathcal{Z}} \theta d Q \in C, \quad Q \in L_{\lambda_{\diamond}^{*}} R \tag{C}
\end{equation*}
$$

where $C$ is a convex subset of $\mathcal{X}_{o}$.
2.3. The extended entropy minimization problem $\left(\overline{\mathrm{P}}_{C}\right)$. The extended entropy is defined by

$$
\begin{equation*}
\bar{I}(\ell)=I\left(\ell^{a}\right)+I^{s}\left(\ell^{s}\right), \quad \ell \in L_{\lambda_{\diamond}}^{\prime} \tag{2.7}
\end{equation*}
$$

where, using the notation of Theorem 2.2,

$$
I^{s}\left(\ell^{s}\right)=\iota_{\mathrm{dom} I_{\gamma}}^{*}\left(\ell^{s}\right)=\sup \left\{\left\langle\ell^{s}, u\right\rangle ; u \in L_{\lambda_{\diamond}}, I_{\gamma}(u)<\infty\right\} \in[0, \infty]
$$

The associated extended minimization problem is

$$
\begin{equation*}
\text { minimize } \bar{I}(\ell) \text { subject to }\langle\theta, \ell\rangle \in C, \quad \ell \in L_{\lambda_{\diamond}}^{\prime} \tag{P}
\end{equation*}
$$

$\bar{I}$ is the greatest convex $\sigma\left(L_{\lambda_{\diamond}}^{\prime}, L_{\lambda_{\diamond}}\right)$-lower semicontinuous extension of $I$ to $L_{\lambda_{\diamond}}^{\prime} \supset L_{\lambda_{\odot}^{*}} R$, see 15 for references about this result.
2.4. Good and bad constraints. If the Young function $\lambda_{\diamond}$ doesn't satisfy the $\Delta_{2^{-}}$ condition (see Definition 2.4), for instance if it has an exponential growth at infinity as in (3.1) below, the small Orlicz space $E_{\lambda_{\diamond}}$ may be a proper subset of $L_{\lambda_{\odot}}$. Consequently, for some functions $\theta$, the integrability property

$$
\begin{equation*}
\left\langle\mathcal{Y}_{o}, \theta(\cdot)\right\rangle \subset E_{\lambda_{0}} \tag{2.8}
\end{equation*}
$$

or equivalently

$$
\forall y \in \mathcal{Y}_{o}, \int_{\mathcal{Z}} \lambda(z,\langle y, \theta(z)\rangle) R(d z)<\infty
$$

may not be satisfied while the weaker property $(2.6):\left\langle\mathcal{Y}_{o}, \theta(\cdot)\right\rangle \subset L_{\lambda_{\diamond}}$, or equivalently

$$
\forall y \in \mathcal{Y}_{o}, \exists \alpha>0, \int_{\mathcal{Z}} \lambda(z, \alpha\langle y, \theta(z)\rangle) R(d z)<\infty
$$

holds. In this situation, analytical complications occur (see Section 4). This is the reason why constraint satisfying $\left(A_{\theta}^{\nabla}\right)$ are called good constraints, while constraints satisfying $\left(\widehat{A_{\theta}}\right)$ but not $\left(\widehat{A_{\theta}}\right)$ are called bad constraints.

One denotes $\mathcal{Y}_{L}$ the subset of $\mathcal{X}_{o}^{*}$ which is isomorphic to the strong closure of the subspace $\left\{\langle y, \theta\rangle ; y \in \mathcal{Y}_{o}\right\}$ in $L_{\lambda_{\odot}}$. Under the assumption $\left(\overline{A_{\theta}^{\nabla}}\right)$, for all $y \in \mathcal{Y}_{o},\langle y, \theta\rangle \in E_{\lambda_{\circ}}$ and $\mathcal{Y}_{L}=\mathcal{Y}_{E}$ : the strong closure of $\left\{\langle y, \theta\rangle ; y \in \mathcal{Y}_{o}\right\}$ in $E_{\lambda_{\odot}}$.

Under the assumption $\left(A_{\theta}^{\nabla}\right)$, the convex set $C$ is said to be a good constraint set if

$$
\begin{equation*}
T^{-1} C \cap L_{\lambda_{\circ}^{*}} R=\bigcap_{y \in Y}\left\{f R \in L_{\lambda_{\circ}^{*}} R ; \int_{\mathcal{Z}}\langle y, \theta\rangle f d R \geq a_{y}\right\} \text { for some } Y \subset \mathcal{Y}_{E} \tag{2.9}
\end{equation*}
$$

and some function $y \in Y \mapsto a_{y} \in \mathbb{R}$. Under the assumption $\left(A_{\theta}^{\vec{\theta}}\right)$, it is said to be a bad constraint set if

$$
\begin{equation*}
T^{-1} C=\bigcap_{y \in Y}\left\{\ell \in L_{\lambda_{\diamond}}^{\prime} ;\langle\langle y, \theta\rangle, \ell\rangle \geq a_{y}\right\} \text { for some } Y \subset \mathcal{Y}_{L} \tag{2.10}
\end{equation*}
$$

These special shapes (2.9) and (2.10) imply some closure property of $C$. For comparison, if $\left(A_{\theta}^{3}\right)$ holds and $C$ is only supposed to be convex, $T^{-1} C=\bigcap_{(y, a) \in A}\left\{\ell \in L_{\lambda_{\odot}}^{\prime} ;\langle\langle y, \theta\rangle, \ell\rangle>a\right\}$ with $A \subset \mathcal{Y}_{L} \times \mathbb{R}$.
2.5. Several function spaces and cones. To state the extended dual problems ( $\widetilde{\mathrm{D}}_{\mathrm{C}}$ ) and $\left(\overline{\mathrm{D}_{C}}\right)$ below, notation is needed. If $\lambda$ is not an even function, one has to consider

$$
\left\{\begin{array}{l}
\lambda_{+}(z, s)=\lambda(z,|s|)  \tag{2.11}\\
\lambda_{-}(z, s)=\lambda(z,-|s|)
\end{array}\right.
$$

which are Young functions and the corresponding Orlicz spaces.
Let $E$ be a Riesz vector space for the order relation $\leq$. Any $e \in E$ admits a positive part: $e_{+}:=e \vee 0$ and a negative part: $e_{-}:=(-e) \vee 0$. One writes $|e|=e_{+}+e_{-}$. There is a natural order on the algebraic dual space $E^{*}$ of a Riesz vector space $E$ which is defined by: $e^{*} \leq f^{*} \Leftrightarrow\left\langle e^{*}, e\right\rangle \leq\left\langle f^{*}, e\right\rangle$ for any $e \in E$ with $e \geq 0$. A linear form $e^{*} \in E^{*}$ is said to be relatively bounded if for any $f \in E, f \geq 0$, we have $\sup _{e:|e| \leq f}\left|\left\langle e^{*}, e\right\rangle\right|<+\infty$. Although $E^{*}$ may not be a Riesz space in general, the vector space $E^{b}$ of all the relatively bounded linear forms on $E$ is always a Riesz space. In particular, the elements of $E^{b}$ admit a decomposition in positive and negative parts $e^{*}=e_{+}^{*}-e_{-}^{*}$.

Definitions 2.12. For any relatively bounded linear form $\zeta$ on $L_{\lambda_{\diamond}}^{\prime}$ i.e. $\zeta \in L_{\lambda_{\odot}}^{\prime b}$, one writes:

- $\zeta \in K_{\lambda}^{\prime \prime}$ to specify that $\zeta_{ \pm \mid L_{\lambda_{ \pm}}^{\prime} \cap L_{\lambda_{0}}^{\prime}} \in L_{\lambda_{ \pm}}^{\prime \prime}$
- $\zeta \in K_{\lambda^{*}}^{\prime}$ to specify that $\zeta_{ \pm \mid L_{\lambda_{ \pm}^{*}} R \cap L_{\lambda_{*}}^{\prime}} \in L_{\lambda_{ \pm}^{*}}^{\prime}$
- $\zeta \in K_{\lambda}$ to specify that $\zeta_{ \pm \mid L_{\lambda_{ \pm}^{*}} R \cap L_{\lambda_{\diamond}}^{\prime}} \in L_{\lambda_{ \pm}}$
- $\zeta \in K_{\lambda^{*}}^{s}$ to specify that $\zeta_{ \pm \mid L_{\lambda_{ \pm}^{*}} R \cap L_{\lambda_{。}}^{\prime}} \in L_{\lambda_{ \pm}^{*}}^{s}$
- $\zeta \in K_{\lambda}^{s \prime}$ to specify that $\zeta_{ \pm \mid L_{\lambda_{ \pm}}^{s} \cap L_{\lambda_{\theta}}^{\prime}} \in L_{\lambda_{ \pm}}^{s \prime}$
where $\lambda_{ \pm}$are defined at (2.11) and $\zeta_{ \pm \mid L_{ \pm} \cap L_{\lambda_{\odot}}^{\prime}} \in L_{ \pm}^{\prime}$ means that the restriction of $\zeta_{ \pm}$to $L_{ \pm} \cap L_{\lambda_{\diamond}}^{\prime}$ is continuous with respect to relative topology generated by the strong topology of $L_{ \pm}$on $L_{ \pm} \cap L_{\lambda_{\bullet}}^{\prime}$.
(1) The sets $K_{\lambda}^{\prime \prime}, K_{\lambda^{*}}^{\prime}, K_{\lambda}, K_{\lambda^{*}}^{s}$ and $K_{\lambda}^{s \prime}$ are defined to be the corresponding subsets of $L_{\lambda_{\odot}}^{\prime b}$. They are not vector spaces in general but convex cones with vertex 0.
(2) The $\sigma\left(K_{\lambda}^{\prime \prime}, K_{\lambda}^{\prime}\right)$-closure $\bar{A}$ of a set $A$ is defined as follows: $\zeta \in L_{\lambda_{\circ}}^{\prime b}$ is in $\bar{A}$ if $\zeta_{ \pm \mid L_{\lambda_{ \pm}}^{\prime} \cap L_{\lambda_{\odot}}^{\prime}}$ is in the $\sigma\left(L_{\lambda_{ \pm}}^{\prime \prime} \cap L_{\lambda_{\diamond}}^{\prime b}, L_{\lambda_{ \pm}}^{\prime} \cap L_{\lambda_{\odot}}^{\prime}\right)$-closure of $A_{ \pm}=\left\{\zeta_{ \pm} ; \zeta \in A\right\}$. Clearly, $\bar{A}_{ \pm}=\left\{\zeta_{ \pm} ; \zeta \in \bar{A}\right\}$.
One defines similarly the $\sigma\left(K_{\lambda^{*}}^{\prime}, K_{\lambda^{*}}\right), \sigma\left(K_{\lambda}, K_{\lambda}^{\prime}\right), \sigma\left(K_{\lambda^{*}}^{s}, K_{\lambda^{*}}\right)$ and $\sigma\left(K_{\lambda}^{s \prime}, K_{\lambda}^{s}\right)$ closures.
(3) Let $A$ be a subset of $L_{\lambda_{\diamond}}$. Its strong closure $s-c l A$ in $K_{\lambda}$ is the set of all measurable functions $u$ such that $u_{ \pm}$is in the $\|\cdot\|_{\lambda_{ \pm}}$-closure of $A_{ \pm}=\left\{v_{ \pm} ; v \in A\right\}$.

Let $\rho$ be a Young function. By Theorem 2.2, we have $L_{\rho}^{\prime \prime}=\left[L_{\rho} . R \oplus L_{\rho}^{s}\right] \oplus L_{\rho}^{s \prime}$. For any $\zeta \in L_{\rho}^{\prime \prime}=\left(L_{\rho^{*}} R \oplus L_{\rho}^{s}\right)^{\prime}$, let us denote the restrictions $\zeta_{1}=\zeta_{\mid L_{\rho^{*}} R}$ and $\zeta_{2}=\zeta_{\mid L_{\rho}^{s}}$. Since, $\left(L_{\rho^{*}} R\right)^{\prime} \simeq L_{\rho} \oplus L_{\rho^{*}}^{s}$, one sees that any $\zeta \in L_{\rho}^{\prime \prime}$ is uniquely decomposed into

$$
\begin{equation*}
\zeta=\zeta_{1}^{a}+\zeta_{1}^{s}+\zeta_{2} \tag{2.13}
\end{equation*}
$$

with $\zeta_{1}=\zeta_{1}^{a}+\zeta_{1}^{s} \in L_{\rho^{*}}^{\prime}, \zeta_{1}^{a} \in L_{\rho}, \zeta_{1}^{s} \in L_{\rho^{*}}^{s}$ and $\zeta_{2} \in L_{\rho}^{s \prime}$. With our definitions, $K_{\lambda}^{\prime \prime}=$ $\left[K_{\lambda} \oplus K_{\lambda^{*}}^{s}\right] \oplus K_{\lambda}^{s \prime}$ and the decomposition (2.13) holds for any $\zeta \in K_{\lambda}^{\prime \prime}$ with

$$
\left\{\begin{array}{l}
\zeta_{1}=\zeta_{1}^{a}+\zeta_{1}^{s} \in K_{\lambda} \oplus K_{\lambda^{*}}^{s}=K_{\lambda^{*}}^{\prime} \\
\zeta_{2} \in K_{\lambda}^{s \prime}
\end{array}\right.
$$

2.6. The assumptions. Let us first collect the assumptions on $R, \gamma^{*}$ and $\theta$.

Assumptions (A).
$\left(\mathrm{A}_{R}\right)$ It is assumed that the reference measure $R$ is a $\sigma$-finite nonnegative measure on a space $\mathcal{Z}$ endowed with some $R$-complete $\sigma$-field.
$\left(\mathrm{A}_{\gamma^{*}}\right)$ Assumptions on $\gamma^{*}$.
(1) $\gamma^{*}(\cdot, t)$ is $z$-measurable for all $t$ and for $R$-almost every $z \in \mathcal{Z}, \gamma^{*}(z, \cdot)$ is a lower semicontinuous strictly convex $[0,+\infty]$-valued function on $\mathbb{R}$ which attains a unique minimum. Let $m(z)$ denote this minimizer.
(2) It is also assumed that for $R$-almost every $z \in \mathcal{Z}$, the minimum value is $\gamma^{*}(z, m(z))=0$ and there exist $a(z), b(z)>0$ such that $0<\gamma^{*}(z, m(z)+$ $a(z))<\infty$ and $0<\gamma^{*}(z, m(z)-b(z))<\infty$.
(3) $\int_{\mathcal{Z}} \lambda^{*}(z, \alpha m(z)) R(d z)+\int_{\mathcal{Z}} \lambda^{*}(z,-\alpha m(z)) R(d z)<\infty$, for some $\alpha>0$.
(4) For $R$-almost every $z \in \mathcal{Z}, \lim _{t \rightarrow \pm \infty} \gamma_{z}^{*}(t) / t=+\infty$.
$\left(\mathrm{A}_{\theta}\right)$ Assumptions on $\theta$.
(1) for any $y \in \mathcal{Y}_{o}$, the function $z \in \mathcal{Z} \mapsto\langle y, \theta(z)\rangle \in \mathbb{R}$ is measurable;
(2) for any $y \in \mathcal{Y}_{o},\langle y, \theta(\cdot)\rangle=0, R$-a.e. implies that $y=0$;
( $\exists$ ) $\forall y \in \mathcal{Y}_{o}, \exists \alpha>0, \quad \int_{\mathcal{Z}} \lambda(z, \alpha\langle y, \theta(z)\rangle) R(d z)<\infty$.
2.7. The solution of $\left(\mathrm{P}_{C}\right)$. The dual problem associated with $(\mathrm{Pd})$ is

$$
\begin{equation*}
\operatorname{maximize} \inf _{x \in C \cap \mathcal{X}_{L}}\langle y, x\rangle-I_{\gamma}(\langle y, \theta\rangle), \quad y \in \mathcal{Y}_{L} \tag{C}
\end{equation*}
$$

and the extended dual problem associated with $\left(\mathbb{P}_{C}\right)$ is

$$
\begin{equation*}
\operatorname{maximize} \inf _{x \in C \cap \mathcal{X}_{L}}\langle\omega, x\rangle-I_{\gamma}(\langle\omega, \theta\rangle), \quad \omega \in \tilde{\mathcal{Y}} \tag{D}
\end{equation*}
$$

where $\widetilde{\mathcal{Y}}$ is the convex cone of all linear forms $\omega$ on $\mathcal{X}_{o}$ such that $\langle\omega, \theta(\cdot)\rangle_{\mathcal{X}_{o}^{*}, \mathcal{X}_{o}}$ is in the $\sigma\left(K_{\lambda}, K_{\lambda^{*}}\right)$-closure of $\left\{\langle y, \theta\rangle ; y \in \mathcal{Y}_{o}\right\}$.
Let us define

$$
\Gamma^{*}(x)=\sup _{y \in \mathcal{Y}_{o}}\left\{\langle y, x\rangle-\int_{\mathcal{Z}} \gamma(\langle y, \theta\rangle) d R\right\}, \quad x \in \mathcal{X}_{o}
$$

the convex conjugate of

$$
\Gamma(y)=\int_{\mathcal{Z}} \gamma(\langle y, \theta\rangle) d R, \quad y \in \mathcal{Y}_{o}
$$

Theorem 2.14 (Problem $\left(\mathrm{P}_{C}\right)$ ). Suppose that
(1) the assumptions ( $A$ ) and ( $A_{\theta}^{\#}$ ) are satisfied;
(2) $C$ satisfies (2.9).

Then:
(a) The following dual equality for ( Pd ) holds:

$$
\inf \left(\mathrm{P}_{C}\right)=\sup \left(\mathrm{D}_{C}\right)=\sup \left(\widetilde{\mathrm{D}}_{C}\right)=\inf _{x \in C} \Gamma^{*}(x) \in[0, \infty]
$$

(b) If $C \cap \operatorname{dom} \Gamma^{*} \neq \emptyset$ or equivalently $C \cap T \operatorname{dom} I \neq \emptyset$, then $\left(\mathbb{P}_{C}\right)$ admits a unique solution $\widehat{Q}$ in $L_{\lambda_{0}^{*}} R$ and any minimizing sequence $\left(Q_{n}\right)_{n \geq 1}$ converges to $\widehat{Q}$ with respect to the topology $\sigma\left(L_{\lambda_{\theta}^{*}} . R, E_{\lambda_{\circ}}\right)$.
Suppose that in addition $C \cap \operatorname{icordom} \Gamma^{*} \neq \emptyset$ or equivalently $C \cap \operatorname{icor}(T \operatorname{dom} I) \neq \emptyset$.
(c) Let us define $\hat{x} \triangleq \int_{\mathcal{Z}} \theta d \widehat{Q}$ in the weak sense with respect to the duality $\left\langle\mathcal{Y}_{o}, \mathcal{X}_{o}\right\rangle$. There exists $\tilde{\omega} \in \widetilde{\mathcal{Y}}$ such that

$$
\begin{cases}(a) & \hat{x} \in C \cap \operatorname{dom} \Gamma^{*}  \tag{2.15}\\ (b) & \langle\tilde{\omega}, \hat{x}\rangle_{\mathcal{X}_{o}^{*}, \mathcal{X}_{o}} \leq\langle\tilde{\omega}, x\rangle_{\mathcal{X}_{\mathcal{X}}^{*}, \mathcal{X}_{o}}, \forall x \in C \cap \operatorname{dom} \Gamma^{*} \\ (c) & \widehat{Q}(d z)=\gamma_{z}^{\prime}(\langle\tilde{\omega}, \theta(z)\rangle) R(d z) .\end{cases}
$$

Furthermore, $\widehat{Q} \in L_{\lambda_{8}^{*}} R$ and $\tilde{\omega} \in \tilde{\mathcal{Y}}$ satisfy (2.15) if and only if $\widehat{Q}$ solves (P]) and $\tilde{\omega}$ solves ( $\widetilde{\mathrm{D}_{C}}$ ).
(d) Of course, 2.15-c) implies

$$
\begin{equation*}
\hat{x}=\int_{\mathcal{Z}} \theta \gamma^{\prime}(\langle\tilde{\omega}, \theta\rangle) d R \tag{2.16}
\end{equation*}
$$

in the weak sense. Moreover,

1. $\hat{x}$ minimizes $\Gamma^{*}$ on $C$,
2. $I(\widehat{Q})=\Gamma^{*}(\hat{x})=\int_{\mathcal{Z}} \gamma^{*} \circ \gamma^{\prime}(\langle\tilde{\omega}, \theta\rangle) d R<\infty$ and
3. $I(\widehat{Q})+\int_{\mathcal{Z}} \gamma(\langle\tilde{\omega}, \theta\rangle) d R=\int_{\mathcal{Z}}\langle\tilde{\omega}, \theta\rangle d \widehat{Q}$.

Proof. This result is [15, Theorem 5.2].
Following the terminology of Ney [22, 23], as it shares the properties (a), (b) and (2.16), the minimizer $\hat{x}$ is called a dominating point of $C$ with respect to $\Gamma^{*}$ (see Definition 6.2 below).
2.8. The solutions of $\left(\overline{\mathrm{P}}_{C}\right)$. Let us turn our attention to the extended problem $\left(\overline{\mathrm{P}}_{C}\right)$.

Denoting the operator $T: \ell \in L_{\lambda_{\circ}}^{\prime} \mapsto\langle\theta, \ell\rangle \in \mathcal{X}_{o}$, it is shown in [15] that $T L_{\lambda_{0}}^{\prime} \subset \mathcal{X}_{L}$ where $\mathcal{X}_{L} \subset \mathcal{X}_{o}$ is the topological dual space of $\left(\mathcal{Y}_{L},|\cdot|_{L}\right)$ with $|y|_{L}=\|\langle y, \theta(\cdot)\rangle\|_{L_{\lambda_{0}}}$. Hence $T: L_{\lambda_{\circ}}^{\prime} \rightarrow \mathcal{X}_{L}$ and one can define its adjoint $T^{*}: \mathcal{X}_{L}^{*} \rightarrow L_{\lambda_{\circ}}^{\prime *}$ for all $\omega \in \mathcal{X}_{L}^{*}$ by: $\left\langle\ell, T^{*} \omega\right\rangle_{L_{\lambda_{\circ}}^{\prime}, L_{\lambda_{o}}^{*}}=\langle T \ell, \omega\rangle_{\mathcal{X}_{L}, \mathcal{X}_{L}^{*}}, \forall \ell \in L_{\lambda_{o}}^{\prime}$. We have the inclusions $\mathcal{Y}_{o} \subset \mathcal{Y}_{L} \subset \mathcal{X}_{L}^{*}$.
The extended dual problem is

$$
\begin{equation*}
\operatorname{maximize} \inf _{x \in C \cap \mathcal{X}_{L}}\langle\omega, x\rangle-I_{\gamma}\left(\left[T^{*} \omega\right]_{1}^{a}\right)-\iota_{D}\left(\left[T^{*} \omega\right]_{2}\right), \quad \omega \in \overline{\mathcal{Y}} \tag{D}
\end{equation*}
$$

where $D$ is the $\sigma\left(K_{\lambda}^{s \prime}, K_{\lambda}^{s}\right)$-closure of $\operatorname{dom} I_{\gamma}$ and $\overline{\mathcal{Y}}$ is the cone of all $\omega \in \mathcal{X}_{L}^{*}$ such that $T^{*} \omega \in K_{\lambda}^{\prime \prime}$ and $\left[T^{*} \omega\right]_{1}^{s}=0$.
As $R$ is assumed to be $\sigma$-finite, there exists a measurable partition $\left(\mathcal{Z}_{k}\right)_{k \geq 1}$ of $\mathcal{Z}: \bigsqcup_{k} \mathcal{Z}_{k}=$ $\mathcal{Z}$, such that $R\left(\mathcal{Z}_{k}\right)<\infty$ for each $k \geq 1$.

Theorem 2.17 (Problem $\left(\overline{\mathrm{P}}_{C}\right)$ ). Suppose that
(1) the assumptions (A) are satisfied;
(2) $C$ satisfies (2.19);
(3) for each $k \geq 1, L_{\lambda_{\bullet}}\left(\mathcal{Z}_{k}, R_{\mid \mathcal{Z}_{k}}\right)$ is dense in $L_{\lambda_{+}}\left(\mathcal{Z}_{k}, R_{\mid \mathcal{Z}_{k}}\right)$ and $L_{\lambda_{-}}\left(\mathcal{Z}_{k}, R_{\mid \mathcal{Z}_{k}}\right)$ with respect to the topologies associated with $\|\cdot\|_{\lambda_{+}}$and $\|\cdot\|_{\lambda_{-}}$.
Then:
(a) The following dual equality for ( $(\mathbb{P})$ holds:

$$
\inf \left(\overline{\mathrm{P}}_{C}\right)=\inf _{x \in C} \Gamma^{*}(x)=\sup \left(\mathrm{D}_{C}\right)=\sup \left(\overline{\mathrm{D}}_{C}\right) \in[0, \infty]
$$

(b) If $C \cap \operatorname{dom} \Gamma^{*} \neq \emptyset$ or equivalently $C \cap T \operatorname{dom} \bar{I} \neq \emptyset$, then $(\overline{\mathrm{P}})$ admits solutions in $L_{\lambda_{\circ}}^{\prime}$, any minimizing sequence admits $\sigma\left(L_{\lambda_{0}}^{\prime}, L_{\lambda_{0}}\right)$-cluster points and every such point is a solution to ( $(\overline{\mathrm{P}})$.

Suppose that in addition we have

$$
C \cap \text { icordom } \Gamma^{*} \neq \emptyset
$$

or equivalently $C \cap \operatorname{icor}(T \operatorname{dom} \bar{I}) \neq \emptyset$. Then:
(c) Let $\hat{\ell}$ be a solution to $(\overline{\mathrm{P}})$ and denote $\hat{x} \triangleq T \hat{\ell}$. There exists $\bar{\omega} \in \overline{\mathcal{Y}}$ such that

$$
\begin{cases}(a) & \hat{x} \in C \cap \operatorname{dom} \Gamma^{*}  \tag{2.18}\\ (b) & \langle\bar{\omega}, \hat{x}\rangle_{\mathcal{X}^{*}, \mathcal{X}_{L}} \leq\langle\bar{\omega}, x\rangle_{\mathcal{X}_{L}^{*}, \mathcal{X}_{L}}, \forall x \in C \cap \operatorname{dom} \Gamma^{*} \\ (c) & \hat{\ell} \in \gamma_{z}^{\prime}\left(\left[T^{*} \bar{\omega}\right]_{1}^{a}\right) R+D^{\perp}\left(\left[T^{*} \bar{\omega}\right]_{2}\right)\end{cases}
$$

where

$$
D^{\perp}(\eta)=\left\{\ell \in L_{\lambda_{o}}^{s} ; \forall h \in L_{\lambda_{o}}, \eta+h \in D \Rightarrow\langle h, \ell\rangle \leq 0\right\}
$$

is the outer normal cone of $D$ at $\eta$.
$T^{*} \bar{\omega}$ is in the $\sigma\left(K_{\lambda}^{\prime \prime}, K_{\lambda}^{\prime}\right)$-closure of $T^{*}(\operatorname{dom} \Gamma)$ and there exists some $\tilde{\omega} \in \mathcal{X}_{o}^{*}$ such that

$$
\left[T^{*} \bar{\omega}\right]_{1}^{a}=\langle\tilde{\omega}, \theta(\cdot)\rangle_{\mathcal{X}_{o}^{*}, \mathcal{X}_{o}}
$$

is a measurable function in the strong closure of $T^{*}(\operatorname{dom} \Gamma)$ in $K_{\lambda}$.
Furthermore, $\hat{\ell} \in L_{\lambda_{0}}^{\prime}$ and $\bar{\omega} \in \overline{\mathcal{Y}}$ satisfy (2.18) if and only if $\hat{\ell}$ solves $\left(\overline{\mathbb{P}_{C}}\right)$ and $\bar{\omega}$ solves ( $\bar{D}_{C}$ ).
(d) Of course, (2.18-c) implies

$$
\begin{equation*}
\hat{x}=\int_{\mathcal{Z}} \theta \gamma^{\prime}(\langle\tilde{\omega}, \theta\rangle) d R+\left\langle\theta, \hat{\ell}^{s}\right\rangle \tag{2.19}
\end{equation*}
$$

Moreover,

1. $\hat{x}$ minimizes $\Gamma^{*}$ on $C$,
2. $\bar{I}(\hat{\ell})=\Gamma^{*}(\hat{x})=\int_{\mathcal{Z}} \gamma^{*} \circ \gamma^{\prime}(\langle\tilde{\omega}, \theta\rangle) d R+\left\langle T^{*} \bar{\omega}, \hat{\ell}^{s}\right\rangle_{K_{\lambda}^{s}, K_{\lambda}^{s}}<\infty$,
3. $I\left(\hat{\ell}^{a}\right)+\int_{\mathcal{Z}} \gamma(\langle\tilde{\omega}, \theta\rangle) d R=\int_{\mathcal{Z}}\langle\tilde{\omega}, \theta\rangle d \hat{\ell}^{a}$ and
4. $\bar{I}\left(\hat{\ell}^{s}\right)=\sup \left\{\left\langle u, \hat{\ell}^{s}\right\rangle ; u \in \operatorname{dom} I_{\gamma}\right\}=\left\langle T^{*} \bar{\omega}, \hat{\ell}^{s}\right\rangle_{K_{\lambda}^{s}, K_{\lambda}^{s}}$.

Proof. This result is [15, Theorem 4.6] which is proved without assuming $\left(A_{\gamma^{*}}^{4}\right)$.
Proposition 2.20. For the assumption (3) of Theorem 2.1才 to be satisfied, it is enough that one of these conditions holds
(i) $\lambda$ is even or more generally $0<\lim \inf _{t \rightarrow \infty} \frac{\lambda_{+}}{\lambda_{-}}(t) \leq \lim \sup _{t \rightarrow \infty} \frac{\lambda_{+}}{\lambda_{-}}(t)<+\infty$;
(ii) $\lim _{t \rightarrow \infty} \frac{\lambda_{+}}{\lambda_{-}}(t)=+\infty$ and $\lambda_{-}$satisfies the $\Delta_{2}$-condition (2.4).

Proof. This result is [15, Proposition 4.9].
Remark 2.21. Since it is assumed that $x \in$ icordom $\Gamma^{*}$, no infinite force field (see [18] for this notion) enters the dual representation of $\ell_{x}$. If 0 is the left bound of dom $\gamma^{*}$, one has $\gamma^{\prime}(-\infty)=0$ and in case $x \in \operatorname{dom} \Gamma^{*} \backslash$ icordom $\Gamma^{*}$, it may happen that $\left\langle y_{x}, \theta\right\rangle$ takes (formally) the value $-\infty$ on some subset $\mathcal{Z}_{0}$ of $\mathcal{Z}$ so that $\frac{d e_{x}^{a}}{d R}$ vanishes on $\mathcal{Z}_{0}$. A similar phenomenon occurs if dom $\gamma_{z}^{*}$ admits a finite right bound $\kappa_{z}$ on a set $\mathcal{Z}_{\kappa}$ such that $R\left(\mathcal{Z}_{\kappa}\right)>0$. Then, $\gamma_{z}^{\prime}(+\infty)=\kappa_{z}<\infty$ and $\frac{d \ell_{a}^{a}}{d R}(z)=\kappa_{z}$ on $\mathcal{Z}_{\kappa}$. For the details, see [18].

## 3. Some examples

3.1. Some examples of entropies. Important examples of entropies occur in statistical physics, probability theory and mathematical statistics. Among them the relative entropy plays a prominent role.

Relative entropy. The reference measure $R$ is assumed to be a probability measure. The relative entropy of $Q \in M_{\mathcal{Z}}$ with respect to $R \in P_{\mathcal{Z}}$ is

$$
I(Q \mid R)=\left\{\begin{array}{ll}
\int_{\mathcal{Z}} \log \left(\frac{d Q}{d R}\right) d Q & \text { if } Q \prec R \text { and } Q \in P_{\mathcal{Z}} \\
+\infty & \text { otherwise }
\end{array}, \quad Q \in M_{\mathcal{Z}}\right.
$$

It corresponds to $\gamma_{z}^{*}(t)=\left\{\begin{array}{ll}t \log t-t+1 & \text { if } t>0 \\ 1 & \text { if } t=0 \\ +\infty & \text { if } t<0\end{array}, m(z)=1\right.$ and

$$
\begin{equation*}
\lambda_{z}(s)=e^{s}-s-1, \quad s \in \mathbb{R}, z \in \mathcal{Z} \tag{3.1}
\end{equation*}
$$

A variant. Taking the same $\gamma^{*}$ and removing the unit mass constraint gives

$$
H(Q \mid R)=\left\{\begin{array}{ll}
\int_{\mathcal{Z}}\left[\frac{d Q}{d R} \log \left(\frac{d Q}{d R}\right)-\frac{d Q}{d R}+1\right] d R & \text { if } 0 \leq Q \prec R \\
+\infty, & \text { otherwise }
\end{array}, \quad Q \in M_{\mathcal{Z}}\right.
$$

This entropy is the rate function of (1.3) when $\left(W_{i}\right)_{i \geq 1}$ is an iid sequence of Poisson(1) random weights. If $R$ is $\sigma$-finite, it is the rate function of the LDP of normalized Poisson random measures, see [16].

Extended relative entropy. Since $\lambda(s)=e^{s}-s-1$ and $R \in P_{\mathcal{Z}}$ is a bounded measure, we have $\lambda_{\circ}(s)=\tau(s):=e^{|s|}-|s|-1$ and the relevant Orlicz spaces are

$$
\begin{aligned}
L_{\tau^{*}} & =\left\{f: \mathcal{Z} \rightarrow \mathbb{R} ; \int_{\mathcal{Z}}|f| \log |f| d R<\infty\right\} \\
E_{\tau} & =\left\{u: \mathcal{Z} \rightarrow \mathbb{R} ; \forall \alpha>0, \int_{\mathcal{Z}} e^{\alpha|u|} d R<\infty\right\} \\
L_{\tau} & =\left\{u: \mathcal{Z} \rightarrow \mathbb{R} ; \exists \alpha>0, \int_{\mathcal{Z}} e^{\alpha|u|} d R<\infty\right\}
\end{aligned}
$$

The extended relative entropy is defined by

$$
\begin{equation*}
\bar{I}(\ell \mid R)=I\left(\ell^{a} \mid R\right)+\sup \left\{\left\langle\ell^{s}, u\right\rangle ; u, \int_{\mathcal{Z}} e^{u} d R<\infty\right\}, \quad \ell \in \mathcal{E}(\mathcal{Z}) \tag{3.2}
\end{equation*}
$$

where $\ell=\ell^{a}+\ell^{s}$ is the decomposition into absolutely continuous and singular parts of $\ell$ in $L_{\tau}^{\prime}=L_{\tau^{*}} \oplus L_{\tau}^{s}$, and $\mathcal{E}(\mathcal{Z})=\left\{\ell \in L_{\tau}^{\prime} ; \ell \geq 0,\langle\ell, \mathbf{1}\rangle=1\right\}$. Note that $\mathcal{E}(\mathcal{Z})$ depends on $R$ and that for all $\ell \in \mathcal{E}(\mathcal{Z}), \ell^{a} \in P_{\mathcal{Z}} \cap L_{\tau^{*}} R$.
3.2. Some examples of constraints. Let us consider the two standard constraints which are the moment constraints and the marginal constraints.
3.2.1. Moment constraints. Let $\theta=\left(\theta_{k}\right)_{1 \leq k \leq K}$ be a measurable function from $\mathcal{Z}$ to $\mathcal{X}_{o}=$ $\mathbb{R}^{K}$. The moment constraint is specified by the operator

$$
\int_{\mathcal{Z}} \theta d \ell=\left(\int_{\mathcal{Z}} \theta_{k} d \ell\right)_{1 \leq k \leq K} \in \mathbb{R}^{K}
$$

which is defined for each $\ell \in M_{\mathcal{Z}}$ which integrates all the real valued measurable functions $\theta_{k}$. The adjoint operator is

$$
\langle y, \theta\rangle=\sum_{1 \leq k \leq K} y_{k} \theta_{k}, \quad y=\left(y_{1}, \ldots, y_{K}\right) \in \mathbb{R}^{K} .
$$

Marginal constraints. Let $\mathcal{Z}=A \times B$ be a product space, $M_{A B}$ be the space of all bounded signed measures on $A \times B$ and $U_{A B}$ be the space of all measurable bounded functions $u$ on $A \times B$. Denote $\ell_{A}=\ell(\cdot \times B)$ and $\ell_{B}=\ell(A \times \cdot)$ the marginal measures of $\ell \in M_{A B}$. The constraint of prescribed marginal measures is specified by

$$
\int_{A \times B} \theta d \ell=\left(\ell_{A}, \ell_{B}\right) \in M_{A} \times M_{B}, \quad \ell \in M_{A B}
$$

where $M_{A}$ and $M_{B}$ are the spaces of all bounded signed measures on $A$ and $B$. The function $\theta$ which gives the marginal constraint is

$$
\theta(a, b)=\left(\delta_{a}, \delta_{b}\right), a \in A, b \in B
$$

where $\delta_{a}$ is the Dirac measure at $a$. Indeed, $\left(\ell_{A}, \ell_{B}\right)=\int_{A \times B}\left(\delta_{a}, \delta_{b}\right) \ell(d a d b)$.
More precisely, let $U_{A}, U_{B}$ be the spaces of measurable functions on $A$ and $B$ and take $\mathcal{Y}_{o}=U_{A} \times U_{B}$ and $\mathcal{X}_{o}=U_{A}^{*} \times U_{B}^{*}$. Then, $\theta$ is a measurable function from $\mathcal{Z}=A \times B$ to $\mathcal{X}_{o}=U_{A}^{*} \times U_{B}^{*}$ and the adjoint of the marginal operator

$$
\langle\theta, \ell\rangle=\left(\ell_{A}, \ell_{B}\right) \in U_{A}^{*} \times U_{B}^{*}, \quad \ell \in U_{A B}^{*}
$$

where $\left\langle f, \ell_{A}\right\rangle:=\langle f \otimes 1, \ell\rangle$ and $\left\langle g, \ell_{B}\right\rangle:=\langle 1 \otimes g, \ell\rangle$ for all $f \in U_{A}$ and $g \in U_{B}$, is given by

$$
\begin{equation*}
\langle(f, g), \theta\rangle=f \oplus g \in U_{A B}, \quad f \in U_{A}, g \in U_{B} \tag{3.3}
\end{equation*}
$$

where $f \oplus g(a, b):=f(a)+g(b), a \in A, b \in B$.

## 4. Solving ( $\mathrm{P}_{C}$ ) with bad constraints

In this section, the minimization problem $(\mathbb{P})$ is considered when the constraint func-
 Problem ( $\mathbb{P}_{d}$ ) may not be attained anymore. Nevertheless, minimizing sequences may admit a limit in some sense. As will be seen at Section $\frac{5}{5}$, this phenomenon is tightly linked to the notion of generalized entropic projection introduced by Csiszár.
4.1. The results about minimizing sequences. One starts this section stating its main results at Theorems 4.1 and 4.7 .

Theorem 4.1. The hypotheses of Theorem 2.17 are assumed.
(a) Suppose that $C \cap \operatorname{dom} \Gamma^{*} \neq \emptyset$. Then, the minimization problem $(\mathbb{P} d)$ is attained in $L_{\lambda_{\circ}}^{\prime}$ and all its solutions share the same unique absolutely continuous part $Q_{\diamond} \in L_{\lambda_{0}^{*}} R$.
(b) Suppose that $C \cap$ icordom $\Gamma^{*} \neq \emptyset$. Then, ( $\left.\widetilde{\widetilde{\mathrm{D}}_{C}}\right)$ is attained in $\widetilde{\mathcal{Y}}$ and

$$
\begin{equation*}
Q_{\diamond}(d z)=\gamma_{z}^{\prime}\left(\left\langle\omega_{\diamond}, \theta(z)\right\rangle\right) R(d z) \tag{4.2}
\end{equation*}
$$

where $\omega_{\diamond} \in \widetilde{\mathcal{Y}}$ is any solution to $\left(\widetilde{\mathbb{D}}_{C}\right)$. More, for each such $\omega_{\diamond},\left\langle\omega_{\diamond}, \theta(\cdot)\right\rangle$ is in the strong closure of $T^{*} \operatorname{dom} \Gamma$ in $K_{\lambda}$ and there exists $\bar{\omega} \in \overline{\mathcal{Y}}$ solution of ( $\bar{D}_{C}$ ) such that $\left[T^{*} \bar{\omega}\right]_{1}^{a}=\left\langle\omega_{\diamond}, \theta(\cdot)\right\rangle$.

Proof. • Proof of (a). The attainment statement is Theorem 2.17-b. Let us show that as $\gamma^{*}$ is strictly convex, if $k_{*}$ and $\ell_{*}$ are two solutions of $\left(\mathbb{P}_{C}\right)$, their absolutely continuous parts match:

$$
\begin{equation*}
k_{*}^{a}=\ell_{*}^{a} . \tag{4.3}
\end{equation*}
$$

$k_{*}, \ell_{*}$ are in the convex set $\left\{\ell \in L_{\lambda_{\odot}}^{\prime} ; T \ell \in C\right\}$ and $\inf \left(\overline{\mathrm{P}}_{C}\right)=\bar{I}\left(k_{*}\right)=\bar{I}\left(\ell_{*}\right)$. For all $0 \leq p, q \leq 1$ such that $p+q=1$, as $I$ and $I^{s}$ are convex functions, we have

$$
\begin{aligned}
\inf \left(\overline{\mathrm{P}}_{C}\right) & \leq \bar{I}\left(p k_{*}+q \ell_{*}\right) \\
& =I\left(p k_{*}^{a}+q \ell_{*}^{a}\right)+I^{s}\left(p k_{*}^{s}+q \ell_{*}^{s}\right) \\
& \leq p I\left(k_{*}^{a}\right)+q I\left(\ell_{*}^{a}\right)+p I^{s}\left(k_{*}^{s}\right)+q I^{s}\left(\ell_{*}^{s}\right) \\
& =p \bar{I}\left(k_{*}\right)+q \bar{I}\left(\ell_{*}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)
\end{aligned}
$$

It follows that $I\left(p k_{*}^{a}+q \ell_{*}^{a}\right)+I^{s}\left(p k_{*}^{s}+q \ell_{*}^{s}\right)=p I\left(k_{*}^{a}\right)+q I\left(\ell_{*}^{a}\right)+p I^{s}\left(k_{*}^{s}\right)+q I^{s}\left(\ell_{*}^{s}\right)$. Suppose that $k_{*}^{a} \neq \ell_{*}^{a}$. As $I$ is strictly convex, with $0<p<1$, one gets: $I\left(p k_{*}^{a}+q \ell_{*}^{a}\right)<p I\left(k_{*}^{a}\right)+q I\left(\ell_{*}^{a}\right)$ and this implies that $I^{s}\left(p k_{*}^{s}+q \ell_{*}^{s}\right)>p I^{s}\left(k_{*}^{s}\right)+q I^{s}\left(\ell_{*}^{s}\right)$ which is impossible since $I^{s}$ is convex. This proves (4.3).

- Proof of (b). Let $\bar{\ell}$ be any solution to $\left(\overline{\mathrm{P}}_{C}\right)$. Denoting $\bar{x}^{a}=T \bar{\ell}^{a}$ and $\bar{x}^{s}=T \bar{\ell}^{s}$ one sees with (2.18) that

$$
\left\{\begin{array}{l}
(a) \quad \bar{x}^{a} \in\left[C-\bar{x}^{s}\right] \cap \operatorname{dom} \Gamma^{*} \\
(b) \\
\left\langle\bar{\omega}, \bar{x}^{a}\right\rangle \leq\langle\bar{\omega}, x\rangle, \forall x \in\left[C-\bar{x}^{s}\right] \cap \operatorname{dom} \Gamma^{*} \\
(c) \\
\bar{\ell}^{a}=\gamma_{z}^{\prime}(\langle\tilde{\omega}, \theta\rangle) R
\end{array}\right.
$$

By Theorem 2.17-c, this implies that $\tilde{\omega}$ solves $\left(\widetilde{\mathrm{D}}_{C-\bar{x}^{s}}\right)$.
It remains to show that $\tilde{\omega}$ also solves $\left(\widetilde{\mathrm{D}}_{C}\right)$. Thanks to Theorem 2.17, we have: $\inf _{x \in C}\langle\bar{\omega}, x\rangle=$ $\left\langle T^{*} \bar{\omega}, \bar{\ell}\right\rangle=\left\langle\langle\tilde{\omega}, \theta\rangle, \bar{\ell}^{a}\right\rangle+\bar{I}\left(\bar{\ell}^{s}\right)$ and $\inf _{x \in C}\langle\bar{\omega}, x\rangle-I_{\gamma}(\langle\tilde{\omega}, \theta\rangle)=\sup \left(\overline{\mathrm{D}}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)=$ $I\left(\bar{\ell}^{a}\right)+\bar{I}\left(\bar{\ell}^{s}\right)=\inf \left(\overline{\mathrm{P}}_{C-\bar{x}^{s}}\right)+\bar{I}\left(\bar{\ell}^{s}\right)=\sup \left(\overline{\mathrm{D}}_{C-\bar{x}^{s}}\right)+\bar{I}\left(\bar{\ell}^{s}\right)$. Therefore $\inf _{x \in C}\langle\bar{\omega}, x\rangle=$ $\inf _{x \in C-\bar{x}^{s}}\langle\tilde{\omega}, x\rangle+\bar{I}\left(\bar{\ell}^{s}\right)$ and subtracting $\bar{I}\left(\bar{\ell}^{s}\right)$ from $\left(\overline{\mathrm{D}}_{C}\right)$, one sees that $\tilde{\omega}$ which solves $\left(\widetilde{\mathrm{D}}_{C-\bar{x}^{s}}\right)$ also solves $\left(\widetilde{\mathrm{D}}_{C}\right)$. One completes the proof of the proposition, taking $\omega_{\diamond}=\tilde{\omega}$.
Remark 4.4. Replacing $\bar{\ell}^{s}$ with $t \bar{\ell}^{s}$, the same proof shows that $\tilde{\omega}$ solves $\left(\widetilde{\mathrm{D}}_{C+(t-1) \bar{x}^{s}}\right)$ for any $t \geq 0$.

From now on, one denotes $Q_{\diamond} \in L_{\lambda_{\circ}^{*}} R$ the absolutely continuous part shared by all the solutions of $\left(\overline{\mathbb{P}_{C}}\right)$. Let us introduce

$$
\begin{align*}
\mathcal{C} & =\left\{Q \in L_{\lambda_{0}^{*}} R ; T Q:=\int_{\mathcal{Z}} \theta d Q \in C\right\}  \tag{4.5}\\
\overline{\mathcal{C}} & =\left\{\ell \in L_{\lambda_{0}}^{\prime} ; T \ell:=\langle\theta, \ell\rangle \in C\right\}
\end{align*}
$$

the constraint sets $T^{-1} C \cap L_{\lambda_{8}^{*}} R$ and $T^{-1} C$ on which $I$ and $\bar{I}$ are minimized. We have: $\mathcal{C}=\overline{\mathcal{C}} \cap L_{\lambda_{8}^{*}} R$ and $I=\bar{I}+\iota_{L_{\lambda_{8}^{*}} R}$. Hence, $\inf \left(\overline{\mathrm{P}}_{C}\right) \leq \inf \left(\mathrm{P}_{C}\right)$. Clearly, $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset \Leftrightarrow$ $\inf \left(\mathrm{P}_{C}\right)<\infty$ implies $\overline{\mathcal{C}} \cap \operatorname{dom} \bar{I} \neq \emptyset \Leftrightarrow \inf \left(\overline{\mathrm{P}}_{C}\right)<\infty \Leftrightarrow C \cap \operatorname{dom} \Gamma^{*} \neq \emptyset$.
Of course, if $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset,(\mathbb{P})$ admits nontrivial minimizing sequences. Theorem 4.7 below gives some details about them.

The present paper is concerned with sets $\mathcal{C}$ of the form (4.5), but this is not a restriction as explained in the following remarks.

## Remarks 4.6.

(1) Taking $T^{*}$ to be the identity on $\mathcal{Y}_{o}=E_{\lambda_{\circ}}$ or $L_{\lambda_{0}}$ (being careless with a.e. equality, this corresponds to $\theta(z)$ to be the Dirac measure $\delta_{z}$ ), one sees that ( $\left.A_{\theta}^{\#}\right)$ or ( $\left.A_{A}^{\exists}\right)$ is satisfied respectively. Hence, with $C=\mathcal{C},(\widehat{\mathrm{Pd}})$ is $(\overline{\mathcal{P c}})$.
Consequently, the specific form with $\theta$ and $C$ adds details to the description of $\mathcal{C}$ without loss of generality.
(2) With $\theta, \mathcal{Y}_{o}$ and $C$ as in (1), the assumption $\left(\mathrm{A}_{C}\right)$ is:
(a) Under ( $\left.\overline{A_{\theta}^{\nabla}}\right): \mathcal{C}$ is $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\circ}}\right)$-closed.

Note that if $\lambda_{\diamond}$ and $\lambda_{\diamond}^{*}$ both satisfy the $\Delta_{2}$-condition, as $\mathcal{C}$ is convex, this is equivalent to $\mathcal{C}$ is $\|\cdot\|_{\lambda_{8}^{*}}$-closed.
(b) Under ( $\left.A_{\theta}^{\exists}\right): \mathcal{C}$ is $\sigma\left(L_{\lambda_{8}^{*}} R, L_{\lambda_{\circ}}\right)$-closed.

Note that if $\lambda_{\diamond}^{*}$ satisfies the $\Delta_{2}$-condition, as $\mathcal{C}$ is convex, this is equivalent to $\mathcal{C}$ is $\|\cdot\|_{\lambda_{\rho}^{*}-\text {-closed. }}$

We denote $\|\cdot\|_{\lambda_{o}^{*}}-\operatorname{int}(\mathcal{C})$ the interior of $\mathcal{C}$ in $L_{\lambda_{0}^{*}} R$ with respect to the strong topology of $L_{\lambda_{0}^{*}}$.

Theorem 4.7 (Minimizing sequences of $(\mathbb{P d})$ ). Suppose that the assumptions of Theorem 2.17 are satisfied and $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset$.

Let us consider the following additional conditions.
(1) $a$ - There are finitely many moment constraints, i.e. $\mathcal{X}_{o}=\mathbb{R}^{K}$ (see Section 3.2.1) b- $\mathcal{C} \cap$ icordom $I \neq \emptyset$.
(2) $\mathcal{C}$ is a $\sigma\left(L_{\lambda_{0}^{*}} R, L_{\lambda_{\circ}}\right)$-closed convex set with a nonempty $\|\cdot\|_{\lambda_{0}^{*}}$-interior.

Under one of these additional conditions (1) or (2), any minimizing sequence of ( $\mathbb{P} C$ ) converges to $Q_{\diamond}$ with respect to $\sigma\left(L_{\lambda_{8}^{*}} R, E_{\lambda_{\circ}}\right)$ and $I\left(Q_{\diamond}\right) \leq \inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)$.

Proof. This proof relies on results which are stated and proved in the remainder of the present section. It is shown at Lemma 4.9 that any minimizing sequence of $(\overline{\mathrm{P}})$ converges in the sense of the $\sigma\left(L_{\lambda_{\circ}^{*}} R, E_{\lambda_{\odot}}\right)$-topology to $Q_{\diamond}$, whenever $\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)$. But this equality holds thanks to Lemma 4.19 and

- under condition (1): Lemma 4.21-a;
- under condition (2): Corollary 4.23-b.

Let us have a look at the last inequality. For any $\bar{\ell}=\bar{\ell}^{a}+\bar{\ell}^{s}=Q_{\diamond}+\bar{\ell}^{s}$ minimizer of $(\overline{\mathbb{P}})$ and any $\left(Q_{n}\right)_{n \geq 1}$ minimizing sequence of $(\mathbb{P d})$, one obtains

$$
\begin{aligned}
\inf _{n} I\left(Q_{n}\right) & =\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)=\bar{I}(\bar{\ell}) \\
& =I\left(Q_{\diamond}\right)+I^{s}\left(\bar{\ell}^{s}\right) \\
& \geq I\left(Q_{\diamond}\right)
\end{aligned}
$$

with a strict inequality if $I^{s}\left(\bar{\ell}^{s}\right)>0$.
Remarks 4.8.
(1) As regards condition (1), it is not assumed that $\mathcal{C}$ has a nonempty interior.
(2) As regards condition (2):
(i) Any $\sigma\left(L_{\lambda_{0}^{*}} R, L_{\lambda_{0}}\right)$-closed convex set has the form

$$
\mathcal{C}=\bigcap_{u \in U}\left\{\ell \in L_{\lambda_{0}^{*}} R ;\langle u, \ell\rangle \geq a_{y}\right\}
$$

for some $U \subset L_{\lambda_{0}}$;
(ii) For $\mathcal{C}$ to have a nonempty $\|\cdot\|_{\lambda_{0}^{*}}$-interior, it is enough that $C \cap \mathcal{X}_{L}$ has a nonempty interior in $\mathcal{X}_{L}$ endowed with the uniform dual norm $|\cdot|_{L}^{*}$. This is a consequence of 14, Lemma 4.13-e].
(3) The last quantity $I^{s}\left(\bar{\ell}^{s}\right)=\inf \left(\mathrm{P}_{C}\right)-I\left(Q_{\diamond}\right)$ is precisely the gap of lower $\sigma\left(L_{\lambda_{8}^{*}} R, E_{\lambda_{\odot}}\right)$ semicontinuity of $I: \lim _{n} Q_{n}=Q_{\diamond}$ and $\inf \left(\mathrm{P}_{C}\right)=\liminf _{n} I\left(Q_{n}\right) \geq I\left(\lim _{n} Q_{n}\right)=$ $I\left(Q_{\diamond}\right)$.
4.2. A preliminary lemma. Preliminary results for the proof of Theorem 4.7 are stated below at Lemma 4.9. The assumption ( $A_{\theta}^{3}$ ) about the bad constraint is $T^{*} \mathcal{Y}_{o} \subset L_{\lambda_{0}}$.
Lemma 4.9. Suppose that the assumptions of Theorem 2.17 are satisfied, $\inf \left(\mathrm{P}_{C}\right)<\infty$ and

$$
\begin{equation*}
\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right) \tag{4.10}
\end{equation*}
$$

Then, any minimizing sequence of ( $\left(\overline{P_{C}}\right)$ converges to $Q_{\diamond}$ with respect to $\sigma\left(L_{\lambda_{\theta}^{*}} R, E_{\lambda_{\circ}}\right)$.
Proof. Let $\left(Q_{n}\right)_{n \geq 1}$ be a minimizing sequence of $\left(\widetilde{P_{C}}\right)$. Since it is assumed that $\inf \left(\mathrm{P}_{C}\right)=$ $\inf \left(\overline{\mathrm{P}}_{C}\right),\left(Q_{n}\right)_{n \geq 1}$ is also a minimizing sequence of $(\overline{\mathrm{P}})$. With the representation of $\bar{I}$ at (4.11) below, one can apply ([17], Corollary 2.2) which states that $\bar{I}$ is $\sigma\left(L_{\lambda_{\circ}}^{\prime}, L_{\lambda_{\odot}}\right)$-infcompact. Hence, one can extract a $\sigma\left(L_{\lambda_{\rho}}^{\prime}, L_{\lambda_{0}}\right)$-convergent subnet $\left(Q_{\alpha}\right)_{\alpha}$ from $\left(Q_{n}\right)_{n \geq 1}$. Let $\ell_{*} \in \overline{\mathcal{C}}$ denote its limit: we have $\lim _{\alpha} \int_{\mathcal{Z}} u d Q_{\alpha}=\left\langle\ell_{*}, u\right\rangle$ for all $u \in L_{\lambda_{\rho}}$. As $\left\langle\ell_{*}^{s}, u\right\rangle=0$, for all $u \in E_{\lambda_{0}}$ (see Theorem 2.2-a), we obtain: $\lim _{\alpha} \int_{\mathcal{Z}} u d Q_{\alpha}=\int_{\mathcal{Z}} u d \ell_{*}^{a}$ for all $u \in E_{\lambda_{0}}$. This proves that $\left(Q_{\alpha}\right)_{\alpha} \sigma\left(E_{\lambda_{\rho}}^{\prime}, E_{\lambda_{\rho}}\right)$-converges to $\ell_{*}^{a}$. As $E_{\lambda_{\rho}}$ is a separable Banach space ( $L_{\lambda_{\rho}}$ is not separable in general), the topology $\sigma\left(E_{\lambda_{\circ}}^{\prime}, E_{\lambda_{\rho}}\right)=\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\circ}}\right)$ is metrizable and one can extract a convergent subsequence $\left(\tilde{Q}_{k}\right)_{k \geq 1}$ from the convergent net $\left(Q_{\alpha}\right)_{\alpha}$. Hence, $\left(\tilde{Q}_{k}\right)_{k \geq 1}$ is a subsequence of $\left(Q_{n}\right)_{n \geq 1}$ which $\sigma\left(L_{\lambda_{\odot}^{*}} R, E_{\lambda_{\odot}}\right)$-converges to $\ell_{*}^{a}$.
Since $\bar{I}$ is inf-compact, $\ell_{*}$ is a minimizer of $(\mathbb{P C})$ and by Theorem 4.1-a, there is a unique $Q_{\diamond}$ such for any minimizing sequence $\left(Q_{n}\right)_{n \geq 1}, \ell_{*}^{a}=Q_{\diamond}$. Therefore, any convergent subsequence of $\left(Q_{n}\right)_{n \geq 1}$ converges to $Q_{\diamond}$ for $\sigma\left(L_{\lambda_{\circ}^{*}} R, E_{\lambda_{\diamond}}\right)$. As any subsequence of a minimizing sequence is still a minimizing sequence, we have proved that from any subsequence of $\left(Q_{n}\right)_{n \geq 1}$, one can extract a sub-subsequence which converges to $Q_{\diamond}$. This proves that $\left(Q_{n}\right)_{n \geq 1}$ converges to $Q_{\diamond}$ with respect to $\sigma\left(L_{\lambda_{o}^{*}} R, E_{\lambda_{\diamond}}\right)$.
4.3. Sufficient conditions for $\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)$. Our aim now is to obtain sufficient conditions for the identity $\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)$ to hold. Let us rewrite the problems ( $\mathbb{P}_{C}$ ) and $(\overline{\mathrm{P}}$ ) in order to emphasize their differences and analogies. Denote

$$
\begin{aligned}
\Phi_{L}(u) & =I_{\lambda}(u)=\int_{\mathcal{Z}} \lambda(u) d R, \quad u \in L_{\lambda_{0}} \\
\Phi_{E}(u) & =\Phi_{L}(u)+\iota_{E}(u), \quad u \in L_{\lambda_{0}}
\end{aligned}
$$

Their convex conjugates are

$$
\begin{array}{ll}
\Phi_{E}^{*}(\ell)=\sup _{u \in E_{\lambda_{0}}}\left\{\langle\ell, u\rangle-I_{\lambda}(u)\right\}, & \ell \in L_{\lambda_{8}^{*}} R \\
\Phi_{L}^{*}(\ell)=\sup _{u \in L_{\lambda_{o}}}\left\{\langle\ell, u\rangle-I_{\lambda}(u)\right\}, & \ell \in L_{\lambda_{0}}^{\prime}
\end{array}
$$

It is shown in that

$$
\begin{cases}I(\ell)=\Phi_{E}^{*}(\ell-m R), \quad \ell \in E_{\lambda_{\circ}}^{\prime}=L_{\lambda_{*}^{*}} R  \tag{4.11}\\ \bar{I}(\ell)=\Phi_{L}^{*}(\ell-m R), \quad \ell \in L_{\lambda_{o}}^{\prime}=L_{\lambda_{o}^{*}} R \oplus L_{\lambda_{o}}^{s}\end{cases}
$$

Hence, considering the minimization problems

$$
\begin{equation*}
\text { minimize } \Phi_{E}^{*}(\ell) \text { subject to }\langle\theta, \ell\rangle \in C_{o}, \quad \ell \in L_{\lambda_{0}^{*}} R \tag{E}
\end{equation*}
$$

and

$$
\begin{equation*}
\text { minimize } \Phi_{L}^{*}(\ell) \text { subject to }\langle\theta, \ell\rangle \in C_{o}, \quad \ell \in L_{\lambda_{0}}^{\prime} \tag{L}
\end{equation*}
$$

with $C_{o}=C-\langle\theta, m R\rangle$, we see that $\ell_{*}$ is a solution of $\left(\mathbb{\mathbb { P } _ { C }}\right)$ [resp. $\left.(\mathbb{P} d)\right]$ if and only if $\ell_{*}-m R$ is a solution of $\left(\mathrm{P}_{E}\right)$ [resp. $\left.\left(\mathrm{P}_{L}\right)\right]$. It is enough to prove

$$
\begin{equation*}
\inf \left(\mathrm{P}_{E}\right)=\inf \left(\mathrm{P}_{L}\right) \tag{4.12}
\end{equation*}
$$

to get $\inf \left(\mathrm{P}_{C}\right)=\inf \left(\overline{\mathrm{P}}_{C}\right)$.
Basic facts about convex duality. The proof of (4.12) will rely on standard convex duality considerations. Let us recall some related facts, as developed in [26].
Let $A$ and $X$ be two vector spaces, $h: A \rightarrow[-\infty,+\infty]$ a convex function, $T: A \rightarrow X$ a linear operator and $C$ a convex subset of $X$. The primal problem to be considered is the following convex minimization problem

$$
\begin{equation*}
\text { minimize } h(a) \text { subject to } T a \in C, \quad a \in A \tag{P}
\end{equation*}
$$

The primal value-function corresponding to the Fenchel perturbation $F(a, x)=h(a)+$ $\iota_{C}(T a+x), a \in A, x \in X$ is $\varphi(x)=\inf _{a \in A} F(a, x)$, i.e.

$$
\varphi(x)=\inf \{h(a) ; a \in A, T a \in C-x\} \quad x \in X
$$

Denote $A^{*}$ the algebraic dual space of $A$. The convex conjugate of $h$ with respect to the dual pairing $\left\langle A, A^{*}\right\rangle$ is

$$
h^{*}(\nu)=\sup _{a \in A}\{\langle\nu, a\rangle-h(a)\}, \quad \nu \in A^{*} .
$$

Let $Y$ be a vector space in dual pairing with $X$. The adjoint operator $T^{*}: Y \rightarrow A^{*}$ of $T$ is defined for all $y \in Y$ by

$$
\left\langle T^{*} y, a\right\rangle_{A^{*}, A}=\langle y, T a\rangle_{Y, X}, \quad \forall a \in A
$$

The dual problem associated with $(\mathcal{P})$ is

$$
\begin{equation*}
\operatorname{maximize} \inf _{x \in C}\langle y, x\rangle-h^{*}\left(T^{*} y\right), \quad y \in Y \tag{D}
\end{equation*}
$$

Let $U$ be some subspace of $A^{*}$. The dual value-function is

$$
\psi(u)=\sup _{y \in Y}\left\{\inf _{x \in C}\langle y, x\rangle-h^{*}\left(T^{*} y+u\right)\right\}, \quad u \in U
$$

One says that $\langle X, Y\rangle$ is a topological dual pairing if $X$ and $Y$ are topological vector spaces and their topological dual spaces $X^{\prime}$ and $Y^{\prime}$ satisfy $X^{\prime}=Y$ and $Y^{\prime}=X$ up to some isomorphisms.

Theorem 4.13 (Criteria for the dual equality). We assume that $A, U, X$ and $Y$ are locally convex Hausdorff topological vector spaces such that $\langle A, U\rangle$ and $\langle X, Y\rangle$ are topological dual pairings. For the dual equality

$$
\inf (\mathcal{P})=\sup (\mathcal{D})
$$

to hold, it is enough that
(1) (a) $h$ is a convex function and $C$ is a convex subset of $X$,
(b) $\varphi$ is lower semicontinuous at $0 \in X$ and
(c) $\sup (\mathcal{D})>-\infty$
or
(2) (a) $h$ is a convex function and $C$ is a closed convex subset of $X$,
(b) $\psi$ is upper semicontinuous at $0 \in U$ and
(c) $\inf (\mathcal{P})<+\infty$.

Remarks 4.14. About the space $U$.
(a) As regards Criterion (1), the space $U$ is unnecessary.
(b) As regards Criterion (2), it is not assumed that $T^{*} Y \subset U$.

Back to our problem. Let us particularize this framework for the problems $\left(\mathrm{P}_{E}\right)$ and $\left(\mathrm{P}_{L}\right)$. Assuming that $m \equiv 0$, one sees that $\left(\overline{P_{C}}\right)=\left(\mathrm{P}_{E}\right),\left(\overline{\mathrm{P}_{C}}\right)=\left(\mathrm{P}_{L}\right), I=I_{\lambda^{*}}=\Phi_{E}^{*}$, $\bar{I}=\bar{I}_{\lambda^{*}}=\Phi_{L}^{*}, \gamma=\lambda, C=C_{o}$ and so on. This simplifying requirement will be assumed during the proof without loss of generality, see the proof of [15, Theorem 4.6].
Let us first apply the criterion (1) of Theorem 4.13.
Problem ( $\mathrm{P}_{E}$ ) is obtained with $A=L_{\lambda_{0}^{*}} R, X=\mathcal{X}_{L}, Y=\mathcal{Y}_{L}$ equipped with the weak topologies $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$ and $\sigma\left(\mathcal{Y}_{L}, \mathcal{X}_{L}\right)$, and $h=\Phi_{E}^{*}=I$. The corresponding primal valuefunction is

$$
\varphi_{E}(x)=\inf \left\{I(Q) ; \int_{\mathcal{Z}} \theta d Q \in C-x, Q \in L_{\lambda_{0}^{*}} R\right\}, \quad x \in \mathcal{X}_{L}
$$

Under the underlying assumption ( $A_{\theta}$ ) , with [14, Lemma 4.13-g] we have:

$$
\begin{equation*}
T^{*} \mathcal{Y}_{L} \subset L_{\lambda_{0}} \tag{4.15}
\end{equation*}
$$

Hence, one only needs to compute $h^{*}$ on $L_{\lambda_{\odot}} \subset\left[L_{\lambda_{\odot}^{*}} R\right]^{*}=A^{*}$. For each $u \in L_{\lambda_{\odot}}, h^{*}(u)=$


$$
\begin{equation*}
h^{*}(u)=\int_{\mathcal{Z}} \lambda(u) d R, \quad u \in L_{\lambda_{\circ}} \tag{4.16}
\end{equation*}
$$

Therefore, the dual problem associated to $\left(\mathrm{P}_{E}\right)$ is

$$
\begin{equation*}
\operatorname{maximize} \inf _{x \in C}\langle y, x\rangle-\int_{\mathcal{Z}} \lambda\left(T^{*} y\right) d R, \quad y \in \mathcal{Y}_{L} \tag{E}
\end{equation*}
$$

Let us go on with $\left(\mathrm{P}_{L}\right)$. Take $A=L_{\lambda_{\odot}}^{\prime}, X=\mathcal{X}_{L}, Y=\mathcal{Y}_{L}$ equipped with the weak topologies $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$ and $\sigma\left(\mathcal{Y}_{L}, \mathcal{X}_{L}\right)$, and $h=\Phi_{L}^{*}=\bar{I}$. The function $h^{*}$ in restriction to $L_{\lambda_{\rho}}$ is still given by (4.16) since $u \in L_{\lambda_{\circ}} \mapsto \int_{\mathcal{Z}} \lambda(u) d R$ is closed convex (Fatou's lemma). The primal value-function is

$$
\varphi_{L}(x)=\inf \left\{\bar{I}(\ell) ;\langle\theta, \ell\rangle \in C-x, \ell \in L_{\lambda_{0}}^{\prime}\right\}, \quad x \in \mathcal{X}_{L}
$$

and the dual problem associated to $\left(\mathrm{P}_{L}\right)$ is

$$
\left(\mathrm{D}_{L}\right)=\left(\mathrm{D}_{E}\right) .
$$

Lemma 4.17. Suppose that $T^{*} \mathcal{Y}_{L} \subset L_{\lambda_{\circ}}$ and $C \cap \mathcal{X}_{L}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed, then $\varphi_{L}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-lower semicontinuous.
Proof. Defining $\tilde{\varphi}(x):=\varphi_{L}(-x)$ and $\bar{J}(x):=\inf \left\{\bar{I}(\ell) ; \ell \in L_{\lambda_{0}}^{\prime}:\langle\theta, \ell\rangle=x\right\}, x \in \mathcal{X}_{L}$, we obtain that $\tilde{\varphi}$ is the inf-convolution of $\bar{J}$ and the convex indicator of $-C: \iota_{-C}$. That is $\tilde{\varphi}(x)=\left(\bar{J} \square \iota_{-C}\right)(x)=\inf \left\{\bar{J}(y)+\iota_{-C}(z) ; y, z, y+z=x\right\}$.
As already seen, $\bar{I}$ is $\sigma\left(L_{\lambda_{\circ}}^{\prime}, L_{\lambda_{\odot}}\right)$-inf-compact and $T$ is $\sigma\left(L_{\lambda_{\odot}}^{\prime}, L_{\lambda_{\odot}}\right)-\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-continuous, see [14, Lemma 4.13-h]. It follows that $\bar{J}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-inf-compact. As $C \cap \mathcal{X}_{L}$ is assumed to be $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed, $\iota_{-C}$ is lower semicontinuous. Finally, being the inf-convolution of an inf-compact function and a lower semicontinuous function, $\tilde{\varphi}$ is lower semicontinuous, and so is $\varphi_{L}$.

As $\bar{I}$ and $C$ are assumed to be convex and $\sup \left(\mathrm{D}_{L}\right) \geq \inf _{x \in C}\langle 0, x\rangle-\int_{\mathcal{Z}} \lambda\left(T^{*} 0\right) d R=$ $0>-\infty$, this lemma allows us to apply Criterion (1) of Theorem 4.13 to obtain

$$
\begin{equation*}
\inf \left(\mathrm{P}_{L}\right)=\sup \left(\mathrm{D}_{L}\right) \tag{4.18}
\end{equation*}
$$

Since $\bar{I}$ and $I$ match on $L_{\lambda_{0}^{*}} R$, we have $\inf \left(\mathrm{P}_{L}\right) \leq \inf \left(\mathrm{P}_{E}\right)$. Putting together these considerations gives us

$$
\sup \left(\mathrm{D}_{E}\right)=\sup \left(\mathrm{D}_{L}\right)=\inf \left(\mathrm{P}_{L}\right) \leq \inf \left(\mathrm{P}_{E}\right)
$$

Since the desired equality $(\boxed{4.10})$ is equivalent to $\inf \left(\mathrm{P}_{L}\right)=\inf \left(\mathrm{P}_{E}\right)$, we have proved
Lemma 4.19. The equality ( 4.10 ) holds if and only if we have the dual equality

$$
\begin{equation*}
\inf \left(\mathrm{P}_{E}\right)=\sup \left(\mathrm{D}_{E}\right) \tag{4.20}
\end{equation*}
$$

This happens if and only if $\varphi_{E}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-lower semicontinuous at $x=0$.
Let us now give a couple of simple criteria for this property to be realized.

## Lemma 4.21.

(a) Suppose that there are finitely many constraints (i.e. $\mathcal{X}_{o}$ is finite dimensional) and $\mathcal{C} \cap$ icordom $I \neq \emptyset$, then $\varphi_{E}$ is continuous at 0 .
(b) Suppose that $\left(\overline{A_{\theta}^{\nabla}}\right)$ is satisfied and $C$ is $\sigma\left(\mathcal{X}_{E}, \mathcal{Y}_{E}\right)$-closed, then $\varphi_{E}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-lower semicontinuous.

Proof. • Proof of (a). To get (a), remark that a convex function on a finite dimensional space is lower semicontinuous on the intrinsic core of its effective domain. By [14, Lemma 4.13], $T$ is $\sigma\left(L_{\lambda_{\circ}^{*}} R, L_{\lambda_{\odot}}\right)$-continuous and the assumption $\mathcal{C} \cap$ icordom $I \neq \emptyset$ implies that 0 belongs to icordom $\varphi_{E}$.

- Proof of $(\mathrm{b})$. It is similar to the proof of Lemma 4.17. The assumption $\left(A_{\theta}^{\nabla}\right)$ insures that $\mathcal{X}_{E}=\mathcal{X}_{L}$ and $T$ is $\sigma\left(L_{\lambda_{\diamond}^{*}} R, E_{\lambda_{\diamond}}\right)-\sigma\left(\mathcal{X}_{E}, \mathcal{Y}_{E}\right)$-continuous.

It follows from Lemma 4.21-b, Lemma 4.19, the remark at (4.12) and Lemma 4.9 that under the good constraint assumptions $\left(A^{\nabla}\right)$, if $C \cap \operatorname{dom} \Gamma^{*} \neq \emptyset$, then any minimizing sequence of $(\overline{\mathrm{Pd}})$ converges with respect to the topology $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\diamond}}\right)$ to the unique solution $\widehat{Q}$ of $\left(\widehat{\mathrm{P}_{d}}\right)$. This is Theorem 2.14 b .

Using Criterion (2). Up to now, we only used Criterion (1) of Theorem 4.13. In the following lines, we are going to use Criterion (2) to prove (4.20) under additional assumptions.

Let us go back to Problem $\left(\mathrm{P}_{E}\right)$. It is still assumed without loss of generality that $m=0$ and $\gamma=\lambda$. One introduces a space $U$ and a dual value-function $\psi$ on $U$. The framework of Theorem 4.13 is preserved when taking $X=\mathcal{X}_{L}, Y=\mathcal{Y}_{L}$ with the weak topologies $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$ and $\sigma\left(\mathcal{Y}_{L}, \mathcal{X}_{L}\right), h=I$ on $A=L_{\lambda_{\diamond}^{*}} R$ as before and adding the following topological pairing $\langle A, U\rangle$. We endow $A=L_{\lambda_{\diamond}^{*}} R$ with the topology $\sigma\left(L_{\lambda_{\diamond}^{*}} R, L_{\lambda_{\diamond}}\right)$ and take

$$
U=A^{\prime}=\left(L_{\lambda_{\diamond}^{*}} R\right)^{\prime} \simeq L_{\lambda_{\diamond}}
$$

with the topology $\sigma\left(L_{\lambda_{\odot}}, L_{\lambda_{\circ}^{*}}\right)$. By (4.16): $h^{*}=I_{\lambda}$, this leads to the dual value-function

$$
\psi(u)=\sup _{y \in \mathcal{Y}_{L}}\left\{\inf _{x \in C}\langle y, x\rangle-I_{\lambda}\left(T^{*} y+u\right)\right\}, \quad u \in L_{\lambda_{\diamond}}
$$

To apply Criterion (2), let us establish the following
Lemma 4.22. For $\psi$ to be $\sigma\left(L_{\lambda_{\odot}}, L_{\lambda_{\circ}^{*}}\right)$-upper semicontinuous, it is enough that
(a) $T^{*} \mathcal{Y}_{L}$ is a $\sigma\left(L_{\lambda_{\odot}}, L_{\lambda_{\odot}^{*}}\right)$-closed subspace of $L_{\lambda_{\odot}}$ and
(b) the interior of $\mathcal{C}$ in $L_{\lambda_{\diamond}^{*}} R$ with respect to $\|\cdot\|_{\lambda_{\diamond}^{*}}$ is nonempty.

Proof. During this proof, unless specified the topology on $L_{\lambda_{\circ}}$ is $\sigma\left(L_{\lambda_{\odot}}, L_{\lambda_{0}^{*}}\right)$. For all $u \in$ $L_{\lambda_{0}}$,

$$
\begin{aligned}
-\psi(u) & =\inf _{y \in \mathcal{Y}_{L}}\left\{\sup \left\{\left\langle-T^{*} y, \ell\right\rangle ; \ell \in L_{\lambda_{\circ}^{*}} R: T \ell \in C\right\}+I_{\lambda}\left(T^{*} y+u\right)\right\} \\
& =\inf _{v \in V}\left\{\sup \{\langle-v, \ell\rangle ; \ell \in \mathcal{C}\}+I_{\lambda}(v+u)\right\} \\
& =I_{\lambda} \square G(u)
\end{aligned}
$$

where $V=T^{*} \mathcal{Y}_{L}$ and $I_{\lambda} \square G(u)=\inf _{v \in L_{\lambda_{\odot}}}\left\{G(v)+I_{\lambda}(u-v)\right\}$ is the inf-convolution of $I_{\lambda}$ and $G(u)=\iota_{\mathcal{C}}^{*}(u)+\iota_{V}(u), u \in L_{\lambda_{\bullet}}$. Let us show that under the assumption (a),

$$
G=\iota_{\mathcal{C}}^{*}
$$

As $V$ is assumed to be closed, we have $\iota_{V}=\iota_{V \perp}^{*}$ with $V^{\perp}=\left\{k \in L_{\lambda_{0}^{*}} R ;\langle v, k\rangle=0, \forall v \in\right.$ $V\}$. This gives for each $u \in L_{\lambda_{c}^{*}}^{\prime}, G(u)=\iota_{\mathcal{C}}^{*}(u)+\iota_{V_{\perp}}^{*}(u)=\sup _{\ell \in \mathcal{C}}\langle u, \ell\rangle+\sup _{k \in V^{\perp}}\langle u, k\rangle=$ $\sup \left\{\langle u, \ell+k\rangle ; \ell \in \mathcal{C}, k \in V^{\perp}\right\}=\iota_{\mathcal{C}+V^{\perp}}^{*}(u)=\iota_{\mathcal{C}}^{*}(u)$, where the last equality holds since $\mathcal{C}+V^{\perp}=\mathcal{C}$, note that $\operatorname{ker} T=V^{\perp}$.
As convex conjugates, $\iota_{\mathcal{C}}^{*}$ and $I_{\lambda}=h^{*}$ are closed convex functions.
Since for all $u, v \in L_{\lambda_{0}}$,

$$
\begin{aligned}
\iota_{\mathcal{C}}^{*}(v)+I_{\lambda}(u-v) & =\iota_{\mathcal{C}-k}^{*}(v)-\langle v, k\rangle+I_{\lambda}(u-v) \\
& =\iota_{\mathcal{C}-k}^{*}(v)+\langle u-v, k\rangle+I_{\lambda}(u-v)-\langle u, k\rangle,
\end{aligned}
$$

we have

$$
-\psi+\langle\cdot, k\rangle=\iota_{\mathcal{C}-k}^{*} \square\left(I_{\lambda}-\langle\cdot, k\rangle\right)
$$

But, by the assumption (b) there exists some $k$ in $L_{\lambda_{0}^{*}} R$ such that $0 \in \operatorname{int}(\mathcal{C}-k)$. It follows that $\iota_{\mathcal{C}-k}^{*}$ is inf-compact. Finally, $-\psi+\langle\cdot, k\rangle$ is lower semicontinuous, being the inf-convolution of a lower semicontinuous and an inf-compact functions.

## Corollary 4.23.

(a) Assume that $C$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed convex, $T^{*} \mathcal{Y}_{L}$ is a $\sigma\left(L_{\lambda_{\circ}}, L_{\lambda_{0}^{*}}\right)$-closed subspace of $L_{\lambda_{\theta}}, \mathcal{C}$ has a nonempty $\|\cdot\|_{\lambda_{0}^{*}}$-interior and $\inf \left(\mathrm{P}_{E}\right)<\infty$. Then, (4.2Q) is satisfied.
(b) In particular, if $\mathcal{C}$ is $\sigma\left(L_{\lambda_{0}^{*}} R, L_{\lambda_{0}}\right)$-closed convex set with a nonempty $\|\cdot\|_{\lambda_{0}^{*}}$-interior and $\inf \left(\mathrm{P}_{E}\right)<\infty$, then (4.20) is satisfied.

Proof. • Proof of (a). Apply the criterion (2) of Theorem 4.13 with Lemma 4.22.

- Proof of (b). This is (a) with $\mathcal{Y}_{L}=L_{\lambda_{\circ}}$ and $T^{*}=I d$, taking advantage of Remarks 4.6.


## 5. Entropic Projections

The results of the preceding sections are translated in terms of entropic projections.
5.1. Generalized entropic projections. We consider the convex problem

$$
\begin{equation*}
\text { minimize } I(Q) \text { subject to } Q \in \mathcal{C} \tag{C}
\end{equation*}
$$

where $\mathcal{C}$ is a convex subset of $M_{\mathcal{Z}}$.
Let $\left(f_{n}\right)_{n \geq 1}$ be a sequence of measurable functions. Following [4], one says that it converges $R$-loosely in measure to $f$ if for each measurable set $A \in \mathcal{Z}$ such that $R(A)<\infty$ and all $\epsilon>0, \lim _{n \rightarrow \infty} R\left(A \cap\left\{\left|f_{n}-f\right|>\epsilon\right\}=0\right.$.
It converges $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{0}}\right)$-loosely if for each measurable set $A \in \mathcal{Z}$ such that $R(A)<\infty$, $\mathbf{1}_{A} f_{n}$ converges to $\mathbf{1}_{A} f$ with respect to $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\odot}}\right)$.

It converges $L_{1}$－loosely if for each measurable set $A \in \mathcal{Z}$ such that $R(A)<\infty, \mathbf{1}_{A} f_{n}$ converges to $\mathbf{1}_{A} f$ strongly in $L_{1}(R)$ ．
Definitions 5．1．Let $Q$ and $\left(Q_{n}\right)_{n \geq 1}$ in $M_{\mathcal{Z}}$ be absolutely continuous with respect to $R$ ．
（1）One says that $\left(Q_{n}\right)_{n \geq 1}$ converges $R$－loosely in measure to $Q$ if $\frac{d Q_{n}}{d R}$ converges $R$－ loosely in measure to $\frac{d Q}{d R}$ ．
（2）One says that $\left(Q_{n}\right)_{n \geq 1}$ converges $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{0}}\right)$－loosely to $Q$ in $L_{\lambda_{0}^{*}} R$ if $\frac{d Q_{n}}{d R}$ con－ verges $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\circ}}\right)$－loosely to $\frac{d Q}{d R}$ ．
（3）One says that $\left(Q_{n}\right)_{n \geq 1}$ converges $R$－loosely in variation to $Q$ if $\frac{d Q_{n}}{d R}$ converges $L_{1}$－ loosely to $\frac{d Q}{d R}$ ．
Definition 5.2 （Generalized entropic projection）．［⿴囗十，Csiszár］．Suppose that $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset$ and that any minimizing sequence of the problem（（ $\mathbb{P}$ C $)$ converges $R$－loosely in measure to some $Q_{*} \in M_{\mathcal{Z}}$ ．
This $Q_{*}$ is called the generalized entropic projection of $m R$ on $\mathcal{C}$ with respect to I．It may not belong to $\mathcal{C}$ ．In case $Q_{*}$ is in $\mathcal{C}$ ，it is called the entropic projection of $m R$ on $\mathcal{C}$ ．

Theorem 5.3 （Csiszár，（4］）．Suppose that $\left(\mathrm{A}_{R}\right)$ and $\left(\mathrm{A}_{\gamma^{*}}\right)$ are satisfied．Then，$m R$ has a generalized projection on any convex subset $\mathcal{C}$ of $M_{\mathcal{Z}}$ such that $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset$ ．

In［4］$\gamma^{*}$ doesn＇t depend on the variable $z$ ，but the proof remains unchanged with a $z$－dependence．

Proposition 5．4．Let $Q$ and $\left(Q_{n}\right)_{n \geq 1}$ in $M_{\mathcal{Z}}$ be absolutely continuous with respect to $R$ such that $\left(Q_{n}\right)_{n \geq 1}$ is a minimizing sequence which converges loosely in $R$－measure to $Q$ ． Then，
（a）$\left(Q_{n}\right)_{n \geq 1}$ converges $R$－loosely in variation to $Q$ ；
（b）$\left(Q_{n}\right)_{n \geq 1}$ converges $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\circ}}\right)$－loosely to $Q$ ．
Proof．• Proof of（a）．Because of de la Vallée Poussin＇s theorem（using assumption（ $\mathrm{A}_{\gamma^{*}}^{4}$ ））， any minimizing sequence $\left(Q_{n}\right)_{n \geq 1}$ is such that $\left\{\mathbf{1}_{A} \frac{d Q_{n}}{d R}, n \geq 1\right\}$ is uniformly integrable for each $A$ such that $R(A)<\infty$ ．Therefore，if $\left(Q_{n}\right)_{n \geq 1}$ converges loosely in $R$－measure，it converges $R$－loosely in variation．
－Proof of（b）．It is enough to consider a bounded measure $R$ ．If $\lambda_{\diamond}^{*}$ is a finite function， this follows from［24，Proposition 4．5．6］．If dom $\lambda_{\diamond}^{*}$ is bounded，we have to show the $\sigma\left(L_{\infty}, L_{1}\right)$－convergence of $f_{n}=\frac{d Q_{n}}{d R}$ to $f=\frac{d Q}{d R}$ ．But $\left(f_{n}\right)_{n \geq 1}$ is bounded in $L_{\infty}$ and the result follows．

As a direct consequence of Theorem 2．14，if the constraints are good，the generalized entropic projection is the entropic projection．

Proposition 5．5．Suppose that $\left(\mathrm{A}_{R}\right)$ and $\left(\mathrm{A}_{\gamma^{*}}\right)$ hold， $\mathcal{C}$ is convex and $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{0}}\right)$－closed and $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset$ ．Then，the entropic projection $Q_{*}$ exists and is equal to

$$
Q_{*}=\widehat{Q} \in \mathcal{C}
$$

where $\widehat{Q}$ is the minimizer of $(\mathbb{P})$ which is described at Theorem 2．14．
Proof．This is an easy corollary of Theorem 2．14．
A consequence of Proposition 5.5 is that for any $\sigma\left(L_{\lambda_{8}^{*}} R, E_{\lambda_{\odot}}\right)$－closed convex set $\mathcal{C}$ ，the generalized entropic projection is the entropic projection．This is essentially［4，Thm 3－（iii）］．

Csiszár's proof of Theorem 5.3 is based on a parallelogram identity which allows to show that any minimizing sequence is a Cauchy sequence. This result is general but it doesn't tell much about the nature of $Q_{*}$. Let us give some details on the generalized entropic projections in specific situations.

Theorem 5.6. Suppose that the assumptions of Theorem 2.1才 are satisfied and $\mathcal{C}$ given by (4.5) satisfies $\mathcal{C} \cap \operatorname{dom} I \neq \emptyset$. Let us consider the additional conditions:
(1) $a$ - There are finitely many moment constraints, i.e. $\mathcal{X}_{o}=\mathbb{R}^{K}$ (see Section 3.2.1) b- $\mathcal{C} \cap$ icordom $I \neq \emptyset$;
(2) $\mathcal{C}$ is a $\sigma\left(L_{\lambda_{0}^{*}} R, L_{\lambda_{0}}\right)$-closed convex set with a nonempty $\|\cdot\|_{\lambda_{0}^{*}}$-interior.

Then, under one of the conditions (1) or (2), the generalized entropic projection $Q_{*}$ of $m R$ on $\mathcal{C}$ is

$$
Q_{*}=Q_{\diamond}
$$

the absolutely continuous component described at (4.8) and $I\left(Q_{*}\right) \leq \inf _{\mathcal{C}} I$.
Proof. This is a direct consequence of Theorems 5.3 and 4.7.
5.2. The special case of relative entropy. The relative entropy $I(P \mid R)$ and its extension $\bar{I}(\ell \mid R)$ are described at Section 3. The minimization problem is

$$
\begin{equation*}
\text { minimize } I(P \mid R) \text { subject to } \int_{\mathcal{Z}} \theta d P \in C, \quad P \in P_{\mathcal{Z}} \tag{5.7}
\end{equation*}
$$

and its extension is

$$
\begin{equation*}
\text { minimize } \bar{I}(\ell \mid R) \text { subject to }\langle\theta, \ell\rangle \in C, \quad \ell \in \mathcal{E}(\mathcal{Z}) \tag{5.8}
\end{equation*}
$$

We introduce the Cramér transform of the image law of $R$ by $\theta$ on $\mathcal{X}_{o}$ :

$$
\begin{equation*}
\Xi(x)=\sup _{y \in \mathcal{Y}_{o}}\left\{\langle y, x\rangle-\log \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\} \in[0, \infty], \quad x \in \mathcal{X}_{o} \tag{5.9}
\end{equation*}
$$

Proposition 5.10 (Relative entropy subject to good constraints). Let us assume that $\theta$ satisfies the "good constraint" assumption

$$
\begin{equation*}
\forall y \in \mathcal{Y}_{o}, \int_{\mathcal{Z}} e^{\langle y, \theta(z)\rangle} R(d z)<\infty \tag{5.11}
\end{equation*}
$$

and that $C \cap \mathcal{X}_{E}$ is a $\sigma\left(\mathcal{X}_{E}, \mathcal{Y}_{E}\right)$-closed convex subset of $\mathcal{X}_{E}$.
(a) The following dual equality holds:

$$
\inf \left\{I(P \mid R) ;\langle\theta, P\rangle \in C, P \in P_{\mathcal{Z}}\right\}=\sup _{y \in \mathcal{Y}_{o}}\left\{\inf _{x \in C}\langle y, x\rangle-\log \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\} \in[0, \infty]
$$

(b) Suppose that in addition $C \cap \operatorname{dom} \Xi \neq \emptyset$. Then, the minimization problem (5.7) has a unique solution $\widehat{P}$ in $P_{\mathcal{Z}}, \widehat{P}$ is the entropic projection of $R$ on $\mathcal{C}=\left\{P \in P_{\mathcal{Z}}, \int_{\mathcal{Z}} \theta d P \in\right.$ $C\}$.
(c) Suppose that in addition, $C \cap$ icordom $\Xi \neq \emptyset$, then there exists some linear form $\tilde{\omega}$ on $\mathcal{X}_{o}$ such that $\langle\tilde{\omega}, \theta\rangle$ is measurable and

$$
\left\{\begin{array}{l}
\hat{x} \triangleq \int_{\mathcal{Z}} \theta d \widehat{P} \in C \cap \operatorname{dom} \Xi \\
\langle\tilde{\omega}, \hat{x}\rangle \leq\langle\tilde{\omega}, x\rangle, \forall x \in C \cap \operatorname{dom} \Xi \\
\widehat{P}(d z)=\exp \left(\langle\tilde{\omega}, \theta(z)\rangle-\log \int_{\mathcal{Z}} e^{\langle\tilde{\omega}, \theta\rangle} d R\right) R(d z) .
\end{array}\right.
$$

In this situation, $\hat{x}$ minimizes $\Xi$ on $C, I(\widehat{P} \mid R)=\Xi(\hat{x})$ and

$$
\begin{equation*}
\hat{x}=\int_{\mathcal{Z}} \theta(z) \exp \left(\langle\tilde{\omega}, \theta(z)\rangle-\log \int_{\mathcal{Z}} e^{\langle\tilde{\omega}, \theta\rangle} d R\right) R(d z) \tag{5.12}
\end{equation*}
$$

in the weak sense.
Proposition 5.13 (Relative entropy subject to bad constraints). Let us assume that $\theta$ satisfies the "bad constraint" assumption

$$
\begin{equation*}
\forall y \in \mathcal{Y}_{o}, \exists \alpha>0, \int_{\mathcal{Z}} e^{\alpha|\langle y, \theta(z)\rangle|} R(d z)<\infty \tag{5.14}
\end{equation*}
$$

and that $C \cap \mathcal{X}_{L}$ is a $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed convex subset of $\mathcal{X}_{L}$.
(a) The following dual equality holds:

$$
\inf \{\bar{I}(\ell \mid R) ;\langle\theta, \ell\rangle \in C, \ell \in \mathcal{E}(\mathcal{Z})\}=\sup _{y \in \mathcal{Y}_{o}}\left\{\inf _{x \in C}\langle y, x\rangle-\log \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\} \in[0, \infty]
$$

(b) Suppose that in addition $C \cap \operatorname{dom} \Xi \neq \emptyset$. Then, the minimization problem (5.8) is attained in $\mathcal{E}(\mathcal{Z})$ : the set of minimizers is nonempty, convex and $\sigma\left(L_{\tau}^{\prime}, L_{\tau}\right)$-compact. Moreover, all the minimizers share the same unique absolutely continuous part $P_{\diamond} \in$ $P_{\mathcal{Z}} \cap L_{\tau^{*}} R$ which is the generalized entropic projection of $R$ on $\mathcal{C}$.
(c) Suppose that in addition, one of the following conditions
(1) $\mathcal{X}_{o}=\mathbb{R}^{K}$ and $\mathcal{C} \cap$ icordom $I \neq \emptyset$ or
(2) $\|\cdot\|_{\tau^{*}}-\operatorname{int}(\mathcal{C}) \neq \emptyset$.
is satisfied. Then, there exists a linear form $\tilde{\omega}$ on $\mathcal{X}_{o}$ such that $\langle\tilde{\omega}, \theta\rangle$ is measurable, $\int_{\mathcal{Z}} e^{\langle\hat{\omega}, \theta\rangle} d R<\infty$ and

$$
P_{\diamond}(d z)=\exp \left(\langle\tilde{\omega}, \theta(z)\rangle-\log \int_{\mathcal{Z}} e^{\langle\tilde{\omega}, \theta\rangle} d R\right) R(d z)
$$

In Proposition 5.10, $\hat{x}$ is the dominating point in the sense of Ney (see Definition 6.3) of $C$ with respect to $\Xi$. The representation of $\hat{x}$ has already been obtained for $C$ with a nonempty topological interior in $\mathbb{R}^{d}$ by Ney in [23] and in a Banach space setting by Einmahl and Kuelbs in [8]. The representation of the generalized projection $P_{\diamond}$ is obtained with a very different proof by Csiszár [2] and (3], Thm 3). Proposition 5.13 also extends corresponding results of Kuelbs [13] which are obtained in a Banach space setting.
For more details about the minimizers of (5.8), one can look at (19), Theorem 3.4) where a characterization is obtained under the weakest assumption: $C \cap \operatorname{dom} \Xi \neq \emptyset$.

Proof of Propositions 5.10 and 5.13 . They are direct consequences of Theorems 2.14, 2.17, Proposition 5.5, Theorem 5.6 and Lemma 5.15 below. Note that $\lambda$ which is given at (3.1) satisfies the condition (ii) of Proposition 2.20 .

The following lemma allows to apply the results of the present paper with $\gamma(s)=e^{s}-1$ and the extended constraint $\langle(\mathbf{1}, \theta), \ell\rangle \in\{1\} \times C$ : the first component of the constraint insures the unit mass $\langle\mathbf{1}, \ell\rangle=1$, to obtain results in terms of log-Laplace transform.

Lemma 5.15. For all $x \in \mathcal{X}$,
$\sup _{y \in \mathcal{Y}_{o}}\left\{\langle y, x\rangle-\log \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\}=\sup _{\tilde{y} \in \mathbb{R} \times \mathcal{Y}_{o}}\left\{\langle\tilde{y},(1, x)\rangle-\int_{\mathcal{Z}}\left(e^{\langle\tilde{y},(1, \theta)\rangle}-1\right) d R\right\} \in(-\infty,+\infty]$.

Proof. Using the identity: $-\log b=\sup _{a}\left\{a+1-b e^{a}\right\}$, one gets:

$$
\begin{aligned}
\sup _{y \in \mathcal{Y}_{o}}\left\{\langle y, x\rangle-\log \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\} & =\sup _{a \in \mathbb{R}, y \in \mathcal{Y}_{o}}\left\{\langle y, x\rangle+a+1-e^{a} \int_{\mathcal{Z}} e^{\langle y, \theta\rangle} d R\right\} \\
& =\sup _{a \in \mathbb{R}, y \in \mathcal{Y}_{o}}\left\{\langle(a, y),(1, x)\rangle-\int_{\mathcal{Z}} e^{\langle y, \theta\rangle+a} d R+1\right\} \\
& =\sup _{\tilde{y} \in \mathbb{R} \times \mathcal{Y}_{o}}\left\{\langle\tilde{y},(1, x)\rangle-\int_{\mathcal{Z}}\left(e^{\langle\tilde{y},(1, \theta)\rangle}-1\right) d R\right\}
\end{aligned}
$$

which is the desired result.
Example 5.16. Csiszár's example. Comparing Proposition 5.5 with Theorem 5.6, one may wonder if the $\sigma\left(L_{\lambda_{0}^{*}} R, E_{\lambda_{\odot}}\right)$-closedness of the convex set $\mathcal{C}$ is critical for the existence of an entropic projection. The answer is affirmative. In [6, Example 3.2], Csiszár gives an interesting example where the generalized entropic projection can be explicitly computed in a situation where $(\mathbb{P C})$ is not attained, see also [7, Exercise 7.3.11]. This example is the following one.
Take the probability measure on $\mathcal{Z}=[0, \infty)$ defined by $R(d z)=a_{0} \frac{e^{-z}}{1+z^{3}} d z$ where $a_{0}$ is the normalizing constant, $I$ the relative entropy with respect to $R, \theta: z \in[0, \infty) \mapsto z \in$ $\mathcal{X}=\mathbb{R}$ and $C=[c, \infty)$. This gives $\mathcal{C}=\left\{Q \in L_{\tau^{*}} R ; \int_{[0, \infty)} z Q(d z) \geq c, Q([0, \infty))=1\right\}$. The point is that $\theta$ is in $L_{\tau}(R)$ but not in $E_{\tau}(R)$.
By Theorem 5.13, the generalized projection of $R$ on $\mathcal{C}$ exists and is equal to $P_{y}=$ $a_{y} \frac{e^{(y-1) z}}{1+z^{3}} d z$ for some real $y \leq 1$, where $a_{y}$ is the normalizing constant. If it belongs to $\mathcal{C}$, then $\int_{\mathcal{Z}} z P_{y}(d z) \geq c$. But $\sup _{y \leq 1} \int_{\mathcal{Z}} z P_{y}(d z)=\int_{\mathcal{Z}} z P_{1}(d z)=a_{1} \int_{\mathcal{Z}} \frac{z}{1+z^{3}} d z:=x_{*}<\infty$. Therefore, for any $c>x_{*}$, there are no entropic projection but only a generalized one.
A detailed analysis of this example in terms of singular component is done by Léonard and Najim [20, Proposition 3.9]. This example corresponds to a $\sigma\left(L_{\lambda_{\circ}^{*}} R, L_{\lambda_{\odot}}\right)$-closed convex set $\mathcal{C}$ such that no entropic projection exists but only a generalized one.
More details about this example are given below at Example 6.10.

## 6. Dominating points

The underlying assumptions are Theorem 2.17]'s ones. We are going to investigate some relations between dominating points and entropic projections. In the case where the constraint is good, Theorem 2.14 and Proposition 5.5 state that the generalized entropic projection is the entropic projection $Q_{*}=\widehat{Q}$, the minimizer $\bar{x}$ is the dominating point of $C$, it is related to $Q_{*}$ by the identity:

$$
\begin{equation*}
\bar{x}=\left\langle\theta, Q_{*}\right\rangle \tag{6.1}
\end{equation*}
$$

We now look at the situation where the constraint is bad. As remarked at Example 5.16, the above equality may fail. A necessary and sufficient condition (in terms of the function $\Gamma^{*}$ ) for $\bar{x}$ to satisfy (6.1) is obtained at Theorem 6.8.

Following Ney [22, 23], let us introduce the following definition. A point $\hat{x} \in \mathcal{X}_{o}$ sharing the properties (a), (b) and (2.16) of Theorem 2.14 is called a dominating point.
Definition 6.2 (Dominating point). Let $C \subset \mathcal{X}_{o}$ be a convex set such that $C \cap \mathcal{X}_{L}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed. The point $\hat{x} \in \mathcal{X}_{o}$ is called $a \Gamma^{*}$-dominating point if
(a) $\hat{x} \in C \cap \operatorname{dom} \Gamma^{*}$
(b) there exists some linear form $\tilde{\omega}$ on $\mathcal{X}_{o}$ such that $\langle\tilde{\omega}, \hat{x}\rangle \leq\langle\tilde{\omega}, x\rangle$ for all $x \in C \cap$ $\operatorname{dom} \Gamma^{*}$,
(c) $\langle\tilde{\omega}, \theta(\cdot)\rangle$ is measurable and $\hat{x}=\int_{\mathcal{Z}} \theta(z) \gamma^{\prime}(z,\langle\tilde{\omega}, \theta(z)\rangle) R(d z)$.

In the special case where $\Gamma^{*}$ is replaced by the Cramér transform $\Xi$ defined at (5.9), taking Lemma 5.15 into account, this definition becomes the following one.
Definition 6.3. Let $C \subset \mathcal{X}_{o}$ be a convex set such that $C \cap \mathcal{X}_{L}$ is $\sigma\left(\mathcal{X}_{L}, \mathcal{Y}_{L}\right)$-closed. The point $\hat{x} \in \mathcal{X}_{o}$ is called $a \Xi$-dominating point if
(a) $\hat{x} \in C \cap \operatorname{dom} \Xi$
(b) there exists some linear form $\tilde{\omega}$ on $\mathcal{X}_{o}$ such that $\langle\tilde{\omega}, \hat{x}\rangle \leq\langle\tilde{\omega}, x\rangle$, for all $x \in$ $C \cap \operatorname{dom} \Xi$ and
(c) $\langle\tilde{\omega}, \theta(\cdot)\rangle$ is measurable and $\hat{x}=\int_{\mathcal{Z}} \theta(z) \frac{\exp (\langle\tilde{\omega}, \theta(z)\rangle)}{Z(\tilde{\omega})} R(d z)$ where $Z(\tilde{\omega})$ is the unit mass normalizing constant.
Note that this definition is slightly different from the ones proposed by Ney [23] and Einmahl and Kuelbs [8] since $C$ is neither supposed to be an open set nor to have a non-empty interior and $\hat{x}$ is not assumed to be a boundary point of $C$. The above integral representation (c) is (5.12).

Recall that the extended entropy $\bar{I}$ is given by (2.7): $\bar{I}(\ell)=I\left(\ell^{a}\right)+I^{s}\left(\ell^{s}\right), \ell=\ell^{a}+\ell^{s} \in$ $L_{\lambda_{\circ}}^{\prime}$ and define for all $x \in \mathcal{X}_{o}$,

$$
\begin{aligned}
\bar{J}(x) & \triangleq \inf \left\{\bar{I}(\ell) ; \ell \in L_{\lambda_{o}}^{\prime},\langle\theta, \ell\rangle=x\right\} \\
J(x) & \triangleq \inf \left\{I(\ell) ; \ell \in L_{\lambda_{0}^{*}} R,\langle\theta, \ell\rangle=x\right\} \\
J^{s}(x) & \triangleq \inf \left\{I^{s}(\ell) ; \ell \in L_{\lambda_{o}}^{s},\langle\theta, \ell\rangle=x\right\}
\end{aligned}
$$

Because of the decomposition $L_{\lambda_{\circ}}^{\prime} \simeq L_{\lambda_{0}^{*}} R \oplus L_{\lambda_{\circ}}^{s}$, one obtains for all $x \in \mathcal{X}_{o}$,

$$
\begin{aligned}
\bar{J}(x) & =\inf \left\{I\left(\ell_{1}\right)+I^{s}\left(\ell_{2}\right) ; \ell_{1} \in L_{\lambda_{8}^{*}} R, \ell_{2} \in L_{\lambda_{o}}^{s},\left\langle\theta, \ell_{1}+\ell_{2}\right\rangle=x\right\} \\
& =\inf \left\{J\left(x_{1}\right)+J^{s}\left(x_{2}\right) ; x_{1}, x_{2} \in \mathcal{X}_{o}, x_{1}+x_{2}=x\right\} \\
& =J \square J^{s}(x)
\end{aligned}
$$

where $J \square J^{s}$ is the inf-convolution of $J$ and $J^{s}$. By Theorem 2.17-a, $\bar{J}=\Gamma^{*}$ and if $\bar{J}(x)<\infty$, there exists $\ell_{x} \in L_{\lambda_{\circ}}^{\prime}$ such that $\left\langle\theta, \ell_{x}\right\rangle=x$ and $\bar{J}(x)=\bar{I}\left(\ell_{x}\right)$. Let us define

$$
x^{a}:=\left\langle\theta, \ell_{x}^{a}\right\rangle \quad \text { and } \quad x^{s}:=\left\langle\theta, \ell_{x}^{s}\right\rangle
$$

These definitions make sense since $\ell_{x}^{a}$ is the unique (common) absolutely continuous part of the minimizers of $\bar{I}$ on the closed convex set $\left\{\ell \in L_{\lambda_{0}}^{\prime} ;\langle\theta, \ell\rangle=x\right\}$, see Theorem 4.1-a. Of course, we have

$$
x=x^{a}+x^{s}
$$

and as $\bar{J}(x)=\bar{I}\left(\ell_{x}\right)=I\left(\ell_{x}^{a}\right)+I^{s}\left(\ell_{x}^{s}\right) \geq J\left(x^{a}\right)+J^{s}\left(x^{s}\right) \geq J \square J^{s}(x)=\bar{J}(x)$, one gets the following result.
Proposition 6.4. For all $x \in \operatorname{dom} \bar{J}$, we have:

$$
\bar{J}(x)=J\left(x^{a}\right)+J^{s}\left(x^{s}\right), J\left(x^{a}\right)=I\left(\ell_{x}^{a}\right) \text { and } J^{s}\left(x^{s}\right)=I^{s}\left(\ell_{x}^{s}\right)
$$

Now, let us have a look at the dual equalities. The recession function of $\Gamma^{*}$ is defined for all $x$ by

$$
\widetilde{\Gamma^{*}}(x) \triangleq \lim _{t \rightarrow+\infty} \Gamma^{*}(t x) / t \in(-\infty,+\infty] .
$$

Definition 6.5 (Recessive $x$ ). Let us say that $x$ is recessive for $\Gamma^{*}$ if for some $\delta>0$ and $\xi \in \mathcal{X}_{o}, \Gamma^{*}(x+t \xi)-\Gamma^{*}(x)=t \widetilde{\Gamma^{*}}(\xi)$ for all $t \in(-\delta,+\infty)$. It is said to be non-recessive otherwise.

Proposition 6.6. We have $\bar{J}=\Gamma^{*}$ and $J^{s}=\widetilde{\Gamma^{*}}$. Moreover, $J(x)=\Gamma^{*}(x)$ for all nonrecessive $x \in \mathcal{X}_{o}$.
Proof. We have already noted that $\bar{J}=\Gamma^{*}$ and by [14. Theorem 2.6-a], we get: $J^{s}=$ $\iota_{\operatorname{dom} \Gamma}^{*} \Gamma$ : the support function of dom $\Gamma$. Therefore, it is also the recession function of $\Gamma^{*}$. Hence, we have $\Gamma^{*}=\bar{J}=J \square J^{s}=J \square \widetilde{\Gamma^{*}}$.
Comparing $\Gamma^{*}=J \square \widetilde{\Gamma^{*}}$ with the general identity $\Gamma^{*}=\Gamma^{*} \square \widetilde{\Gamma^{*}}$, one obtains that $J(x)=$ $\Gamma^{*}(x)$, for all non-recessive $x \in \mathcal{X}_{o}$.
Proposition 6.7. For all $x \in \operatorname{dom} \Gamma^{*}$, we have:

$$
\Gamma^{*}(x)=\Gamma^{*}\left(x^{a}\right)+\widetilde{\Gamma^{*}}\left(x^{s}\right)
$$

Moreover, $x$ is non-recessive if and only if $x^{s}=0$. In particular, $x^{a}$ is non-recessive.
Proof. By (2.7), we have $\bar{I}\left(\ell_{x}\right)=I\left(\ell_{x}^{a}\right)+\widetilde{I}\left(\ell_{x}^{s}\right)$ where $\widetilde{I}$ is the recession function of $I$. It follows that $\bar{J}(x)=J\left(x_{a}\right)+J^{s}\left(x^{s}\right)$, since $J\left(x_{a}\right)=I\left(\ell_{x}^{a}\right)$ (Proposition 6.4) and the recession function of $\bar{J}$ is $J^{s}$. To show this, note that (see [27])

- $I^{s}$ is the recession function of $\bar{I}$,
- the epigraph of $x \mapsto \inf \{f(\ell) ; \ell, T \ell=x\}$ (with $T$ a linear operator) is "essentially" a linear projection of the epigraph of $f$, (let us call it an inf-projection)
- the epigraph of the recession function is the recession cone of the epigraph and
- the inf-projection of a recession cone is the recession cone of the inf-projection.

The first result now follows from $\bar{J}=\Gamma^{*}$. The same set of arguments also yields the second statement.

Theorem 6.8. Let us assume that the hypotheses of Theorem 2.17 hold and that $C \cap$ icordom $\Gamma^{*} \neq \emptyset$.
(a) Then, a minimizer $\bar{x}$ of $\Gamma^{*}$ on the set $C$ is a $\Gamma^{*}$-dominating point of $C$ if and only if $\bar{x}$ is non-recessive. This is also equivalent to the following statement: "all the solutions of the minimization problem $(\overline{\mathrm{P}}$ ) are absolutely continuous with respect to $R$." In such a case the solution of $\left(\overline{P_{C}}\right)$ is unique and it matches the solution of $\left(\mathbb{P}_{C}\right)$.
(b) In particular when $\Gamma^{*}$ admits a degenerate recession function, i.e. $\widetilde{\Gamma^{*}}(x)=+\infty$ for all $x \neq 0$, then the minimizer $\bar{x}$ is a $\Gamma^{*}$-dominating point of $C$.
(c) The same statements hold with $\Gamma^{*}$ replaced by $\Xi$.

Proof. This is a direct consequence of Theorem 2.17, Proposition 6.7 and Lemma 5.15.
Remark 6.9. A remark about the steepness of the log-Laplace transform. In [13, Thm 1], with the setting of Section 5.2 where $\mathcal{X}_{o}$ is a Banach space, Kuelbs proves a result that is slightly different from statement (b) of the above theorem. It is proved that the existence of a $\Xi$-dominating point for all convex sets $C$ with a nonempty topological interior is equivalent to some property of the Gâteaux derivative of the $\log$-Laplace transform $y \in$ $\mathcal{X}_{o}^{\prime} \mapsto \log \int_{\mathcal{X}_{o}} \exp (\langle y, x\rangle) R \circ \theta^{-1}(d x)$ on the boundary of its domain. This property is an infinite dimensional analogue of the steepness of the log-Laplace transform. It turns out that it is equivalent to the following assumption: the Cramér transform $\Xi$ admits a degenerate recession function.

Example 6.10. Csiszár's example continued. We go on with Example 5.16. By Lemma 5.15, $\Gamma^{*}$ is the Cramér transform of $R$ (identifying $(1, x)$ with $\left.x\right)$. This means that $\Gamma^{*}$ is the convex conjugate of the $\log$-Laplace transform $\Gamma(y)=\log \int_{[0, \infty)} a \frac{e^{(y-1) z}}{1+z^{3}} d z, y \in \mathbb{R}$.

Clearly, $\operatorname{dom} \Gamma=(-\infty, 1]$ and $\Gamma^{\prime}\left(1^{-}\right)=\int_{[0, \infty)} z P_{1}(d z):=x_{*}<\infty$. It follows that for all $x \geq x_{*}, \Gamma^{*}(x)-\Gamma^{*}\left(x_{*}\right)=x-x_{*}$. One deduces from this that $\left(x_{*}, \infty\right)$ is a set of recessive points. By Theorem 6.8, they cannot be dominating points. Note also that the $\log$-Laplace transform $\Gamma$ is not steep.

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