# Foundations of Gauge and Perspective Duality 

James V Burke

> Joint work with
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Thank you Mau Nam Nguyen, Hung M. Phan, and Xianfu (Shawn) Wang!!

## Gauge Optimization and Duality

Suppose $\kappa$ and $\rho$ are gauges.

$$
\begin{array}{llll}
\min _{x} & \kappa(x) & \text { s.t. } & \rho(b-A x) \leq \tau, \\
\max _{y} & \langle b, y\rangle-\tau \rho^{\circ}(y) & \text { s.t. } & \kappa^{\circ}\left(A^{T} y\right) \leq 1, \\
\min _{y} & \kappa^{\circ}\left(A^{T} y\right) & \text { s.t. } & \langle b, y\rangle-\tau \rho^{\circ}(y) \geq 1 . \tag{d}
\end{array}
$$

When $\tau=0$, we define $\tau \rho^{\circ}:=\delta_{\text {cldom } \rho^{\circ}}$.

## Minkowski (gauge) functionals and polarity

Let $0 \in C \subset \mathbb{R}^{n}$ be nonempty, closed, and convex. The gauge function for $C$ is given by

$$
\gamma_{C}(x):=\inf \{t \mid 0 \leq t, x \in t C\}
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where the infimum over the empty set is $+\infty$.

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Example: $\|x\|=\gamma_{\mathbb{B}}(x)$ for any norm with unit ball $\mathbb{B}$.
Gauge functions are sublinear, and so by Hörmander,

$$
\gamma_{C}(x)=\sigma_{D}(x):=\sup \{\langle x, y\rangle \mid y \in D\}
$$

where

$$
D=\{z \mid\langle z, x\rangle \leq 1 \forall x \in C\}=: C^{\circ}
$$

and $\sigma_{D}$ is the support function for the set $D$.

## Polar Gauges

Set $\mathcal{U}_{\kappa}:=\{x \mid \kappa(x) \leq 1\}$ and define the polar gauge by

$$
\kappa^{\circ}(y)=\sup \{\langle y, x\rangle \mid \kappa(x) \leq 1\}=\sigma_{\mathcal{U}_{\kappa}}(y)
$$

If $\kappa$ is a norm then $\kappa^{\circ}$ is the corresponding dual norm.

$$
\text { epi } \kappa^{\circ}=\left\{(y,-\lambda):(y, \lambda) \in(\text { epi } \kappa)^{\circ}\right\}
$$

The generalized Hölder inequality

$$
\langle x, y\rangle \leq \kappa(x) \cdot \kappa^{\circ}(y) \quad \forall x \in \operatorname{dom} \kappa, \forall y \in \operatorname{dom} \kappa^{\circ},
$$

is known as the polar-gauge inequality.
In addition, for $\mathcal{H}_{\kappa}:=\{u \mid \kappa(u)=0\}$, we have
$\mathcal{U}_{\kappa}^{\circ}=\mathcal{U}_{\kappa}{ }^{\circ}, \quad \mathcal{U}_{\kappa}^{\infty}=\mathcal{H}_{\kappa}, \quad(\operatorname{dom} \kappa)^{\circ}=\mathcal{H}_{\kappa^{\circ}}, \quad$ and $\quad \mathcal{H}_{\kappa}^{\circ}=\operatorname{cldom} \kappa^{\circ}$.

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$$

When $\tau=0$, we define $\tau \rho^{\circ}:=\delta_{\text {cl dom } \rho^{\circ}}$.

## Feasibility

Primal, Dual Domains:
$\mathcal{F}_{p}:=\{x \mid \rho(b-A x) \leq \tau\} \quad$ and $\quad \mathcal{F}_{d}:=\left\{y \mid\langle b, y\rangle-\tau \rho^{\circ}(y) \geq 1\right\}$.

Feasibilty: $\begin{array}{ll}\text { Primal } & \mathcal{F}_{p} \cap(\operatorname{dom} \kappa) \\ \text { Dual } & A^{T} \mathcal{F}_{d} \cap\left(\operatorname{dom} \kappa^{\circ}\right)\end{array}$

Relative Strict Feasibilty :
Primal ri $\mathcal{F}_{p} \cap($ ridom $\kappa)$
Dual $\quad A^{T}$ ri $\mathcal{F}_{d} \cap\left(\right.$ ridom $\left.\kappa^{\circ}\right)$

Strict Feasibilty :

$$
\begin{array}{ll}
\text { Primal } & \operatorname{int}(\mathcal{F})_{p} \cap(\operatorname{ridom} \kappa) \\
\text { Dual } & A^{T} \operatorname{int}(\mathcal{F})_{d} \cap\left(\operatorname{ridom} \kappa^{\circ}\right)
\end{array}
$$

## Freund (1987), Friedlander-Macedo-Pong (2014)

$$
v_{p}=\min _{\rho(b-A x) \leq \tau} \kappa(x) \quad v_{d}=\min _{\langle b, y\rangle-\tau \rho^{\circ}(y) \geq 1} \kappa^{\circ}\left(A^{T} y\right)
$$

Theorem: (2014)

1. (Weak duality)

If $x$ and $y$ are P-D feasible, then

$$
1 \leq v_{p} v_{d} \leq \kappa(x) \cdot \kappa^{\circ}\left(A^{T} y\right)
$$

2. (Strong duality)

If the dual (resp. primal) is feasible and the primal (resp. dual) is relatively strictly feasible, then $\nu_{p} \nu_{d}=1$ and the gauge dual (resp. primal) attains its optimal value.

## Infimal Projection Duality Theory

Let $F: \mathbb{R}^{n} \times \mathbb{R}^{m} \rightarrow \overline{\mathbb{R}}$ be closed proper convex, and define the following optimal value functions by inf-projection:

$$
p(y):=\inf _{x} F(x, y) \text { and } q(w):=\inf _{z} F^{*}(w, z)
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p(0) \geq p^{* *}(0)=-q(0) \text { always holds }
\end{gathered}
$$

Duality Theory

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## Duality Theory

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$$

1. If $0 \in \operatorname{ri}(\operatorname{dom} p)$, then $p(0)=-q(0)$ and the infimum $q(0)$ is attained, if finite.
Similarly, if $0 \in \operatorname{ri}(\operatorname{dom} q)$, then $p(0)=-q(0)$ and the infimum $p(0)$ is attained, if finite.

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2. The set $\operatorname{argmax}_{z}-F^{*}(0, z)$ is nonempty and bounded if and only if $0 \in \operatorname{int}(\operatorname{dom} p)$ and $p(0)$ is finite, in which case $\partial p(0)=\operatorname{argmax}_{z}-F^{*}(0, z)$.

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4. Optimal solutions are characterized by
$\left.\begin{array}{l}\bar{x} \in \operatorname{argmin}_{x} F(x, 0) \\ \bar{y} \in \operatorname{argmax}_{z}-F^{*}(0, z) \\ F(\bar{x}, 0)=-F^{*}(0, \bar{z})\end{array}\right\} \Longleftrightarrow(0, \bar{z}) \in \partial F(\bar{x}, 0) \Longleftrightarrow(\bar{x}, 0) \in \partial F^{*}(0, \bar{z})$.

## Fenchel-Rockafellar Duality

$$
\begin{aligned}
& F(x, y)=h(A x+y)+g(x) \\
& p(0)=\inf _{x}\{h(A x)+g(x)\} \text { and } p^{\star \star}(0)=\sup _{z}\left\{-h^{*}(z)-g^{*}\left(-A^{*} z\right)\right\}
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A prototype problem:

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\mathcal{P} \quad & \min \|x\|_{1} \\
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\begin{array}{ll}
g(x)=\|x\|_{1}=\delta^{*}\left(x \mid \mathbb{B}_{\infty}\right) & g^{*}(w)=\delta\left(w \mid \mathbb{B}_{\infty}\right) \\
h(y)=\delta\left(y-b \mid \tau \mathbb{B}_{2}\right) & h^{*}(z)=-\langle z, b\rangle+\delta^{*}\left(z \mid \tau \mathbb{B}_{2}\right)=-\langle z, b\rangle+\tau\|z\|_{2}
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$$

$\mathcal{D}_{\mathrm{L}}$

$$
\begin{gathered}
\sup \langle b, z\rangle-\tau\|z\|_{2} \\
\text { s.t. }\left\|A^{T} z\right\|_{\infty} \leq 1 .
\end{gathered}
$$

## Gauge Duality and Sensitivity

$$
v_{p}(y):=\inf _{\mu>0, x}\{\mu \mid \rho(b-A x+\mu y) \leq \tau, \kappa(x) \leq \mu\}
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v_{p}(y):=\inf _{\mu>0, x}\{\mu \mid \rho(b-A x+\mu y) \leq \tau, \kappa(x) \leq \mu\} \\
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\end{gathered}
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or

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$$

Variational framework:

$$
\begin{aligned}
F(w, \lambda, y) & :=-\lambda+\delta_{(\text {epi } \rho) \times \mathcal{U}_{\kappa}}\left(W\left(\begin{array}{c}
w \\
\lambda \\
y
\end{array}\right)\right), \quad W:=\left(\begin{array}{ccc}
-A & b & I \\
0 & \tau & 0 \\
I & 0 & 0
\end{array}\right) \\
F^{*}(w, \lambda, y) & =\delta_{\text {epi } \rho^{\circ}}\binom{y}{-\sigma^{-1}(1+\lambda-\langle b, y\rangle)}+\kappa^{\circ}\left(w+A^{T} y\right)
\end{aligned}
$$

## Gauge Duality and Sensitivity

$$
p(y):=\inf _{w, \lambda} F(w, \lambda, y)
$$

Theorem: The following relationships hold for the gauge primal-dual pair $G_{p}$ and $G_{d}$.
(a) If the primal is relatively strictly feasible and the dual is feasible, then the set of optimal solutions for the dual is nonempty and coincides with

$$
\partial p(0)=\partial\left(-1 / v_{p}\right)(0) .
$$

If it is further assumed that the primal is strictly feasible, then the set of optimal solutions to the dual is bounded.
(b) If the dual is relatively strictly feasible and the primal is feasible, then the set of optimal solutions for the primal is nonempty with solutions $x^{*}=w^{*} / \lambda^{*}$, where

$$
\left(w^{*}, \lambda^{*}\right) \in \partial v_{d}(0,0) \text { and } \lambda^{*}>0 .
$$

If it is further assumed that the dual is strictly feasible, then the set of optimal solutions to the primal is bounded.

## Gauge Duality and Optimality Conditions

Theorem: Suppose both the gauge primal and gauge dual problems are relatively strictly feasible, and the pair $\left(x^{*}, y^{*}\right)$ is primal-dual feasible. Then $\left(x^{*}, y^{*}\right)$ is primal-dual optimal if and only if it satisfies the conditions

$$
\rho\left(b-A x^{*}\right)=\tau \quad \text { or } \quad \rho^{\circ}\left(y^{*}\right)=0 \quad \text { (primal activity) }
$$

$$
\left\langle b, y^{*}\right\rangle-\tau \rho^{\circ}\left(y^{*}\right)=1
$$

(dual activity)
(constraint alignment)

$$
\left\langle x^{*}, A^{T} y^{*}\right\rangle=\kappa\left(x^{*}\right) \cdot \kappa^{\circ}\left(A^{T} y^{*}\right) \quad \text { (objective alignment) }
$$

$$
\left\langle b A x^{*}, y^{*}\right\rangle=\tau \rho^{\circ}\left(y^{*}\right)
$$

By convention, when $\tau=0, \tau \rho^{\circ}:=\delta_{\text {cldom } \rho^{\circ}}$.

## Gauge primal-dual recovery

Corollary: Suppose that the primal-dual pair $\left(\mathrm{G}_{p}\right)$ and $\left(\mathrm{G}_{d}\right)$ are each relatively strictly feasible. If $y^{*}$ is optimal for $\left(\mathrm{G}_{d}\right)$, then for any primal feasible $x$ the following conditions are equivalent:
(a) $x$ is optimal for $\left(\mathrm{G}_{p}\right)$;
(b) $\left\langle x, A^{T} y^{*}\right\rangle=\kappa(x) \cdot \kappa^{\circ}\left(A^{T} y^{*}\right)$ and $b-A x \in \partial\left(\sigma \rho^{\circ}\right)\left(y^{*}\right)$;
(c) $A^{T} y^{*} \in \kappa^{\circ}\left(A^{T} y^{*}\right) \cdot \partial \kappa(x)$ and $b-A x \in \partial\left(\sigma \rho^{\circ}\right)\left(y^{*}\right)$,
where, by convention, $\sigma \rho^{\circ}=\delta_{\text {cldom } \rho^{\circ}}$ when $\sigma=0$, in which case

$$
\partial\left(\sigma \rho^{\circ}\right)\left(y^{*}\right)=N\left(y^{*} \mid \mathcal{H}_{\rho}^{\circ}\right) .
$$

## Gauge primal-dual recovery from the Lagrange dual

## Theorem:

Suppose that the gauge dual $G_{d}$ is relatively strictly feasible and the primal $G_{p}$ is feasible. Let $L_{p}$ denote the Fenchel-Rockafellar dual of $G_{d}$, and let $\nu_{L}$ denote its optimal value. Then

$$
z^{*} \text { is optimal for } L_{p} \Longleftrightarrow z^{*} / \nu_{L} \text { is optimal for } G_{p}
$$

## Perspective Duality

The Perspective Transform

$$
f^{\pi}(x, \mu):= \begin{cases}\mu f\left(\mu^{-1} x\right), & \mu>0 \\ f^{\infty}(x), & \mu=0 \\ +\infty, & \mu<0\end{cases}
$$

where

$$
f^{\infty}(x):=\sup _{z \in \operatorname{dom}(f)}[f(x+z)-f(x)]
$$

is the horizon function of $f$.

$$
h^{\pi}(y, \mu)=\sigma_{\text {epi } h^{*}}((y,-\mu))
$$

## The Perspective-Polar Transform

$$
\begin{aligned}
f^{\sharp}(x, \xi) & :=\left(f^{\pi}\right)^{\circ}(x, \xi) \\
& =\sigma_{\mathrm{epi}\left(f^{*}\right)^{\circ}}(x,-\xi) \\
& =\gamma_{\mathrm{epi}\left(f^{*}\right)}(x,-\xi) \\
& =\inf \{\mu>0 \mid \xi+\langle z, x\rangle \leq \mu f(z), \forall z\}
\end{aligned}
$$

$f^{\sharp}$ is a gauge.

If $f$ is a gauge, then $f^{\sharp}(x, \xi)=f^{\circ}(x)+\delta_{\mathbb{R}_{-}}(\xi)$.

## Perspective duality

Suppose $f: \mathbb{R}^{n} \rightarrow \overline{\mathbb{R}}_{+}$and $g: \mathbb{R}^{m} \rightarrow \overline{\mathbb{R}}_{+}$are closed, convex and nonnegative over their domains.

$$
\begin{array}{lll}
N_{p} & \min _{x} & f(x) \\
\text { s.t. } & g(b-A x) \leq \sigma, \\
N_{d} & \min _{y, \alpha, \mu} f^{\sharp}\left(A^{T} y, \alpha\right) & \text { s.t. } \quad\langle b, y\rangle-\sigma \cdot g^{\sharp}(y, \mu) \geq 1-(\alpha+\mu)
\end{array}
$$

## The Perspective-Polar of a PLQ Penalty

Piecewise linear-quadratic (PLQ) penalties:

$$
g(y):=\sup _{u \in U}\left\{\langle u, y\rangle-\frac{1}{2}\|L u\|_{2}^{2}\right\}, \quad U:=\left\{u \in \mathbb{R}^{l} \mid W u \leq w\right\},
$$

$$
g^{\sharp}(y, \mu)=\delta_{\mathbb{R}_{-}}(\mu)+\max \left\{\gamma_{U}(y),-(1 / 2 \mu)\|L y\|^{2}\right\}
$$

$$
=\delta_{\mathbb{R}_{-}}(\mu)+\max \left\{-(1 / 2 \mu)\|L y\|^{2}, \max _{i=1, \ldots, k}\left\{W_{i}^{T} y / w_{i}\right\}\right\}
$$

where $W_{1}^{T}, \ldots, W_{k}^{T}$ are the rows of $W$.

## The Perspective Duality for PLQ Penalties

Assume $f$ is a gauge and $g$ is a PLQ penalty, then

$$
\begin{array}{ll}
\min _{(y, \mu, \xi)} & f^{\circ}\left(A^{T} y\right) \\
\text { s.t. } & \langle b, y\rangle+\mu-\sigma \xi=1 \\
& W y \leq \xi w,\left\|\left[\begin{array}{c}
2 L y \\
\xi+2 \mu
\end{array}\right]\right\|_{2} \leq \xi-2 \mu
\end{array}
$$

## Perspective Duality Numerics

$$
\begin{array}{ll}
\min _{x} & \|x\|_{1} \\
\text { s.t. } & \sum_{i=1}^{m} V\left((A x-b)_{i}\right) \leq \sigma
\end{array}
$$

where $V$ is the Huber function


Experiment:
$m=120, n=512, \sigma=0.2, \eta=1$, and $A$ is a Gaussian matrix. The true solution $x_{\text {true }} \in\{-1,0,1\}$ is a spike train which has been constructed to have 20 nonzero entries, and the true noise $b-A x_{\text {true }}$ has been constructed to have 5 outliers.

## Perspective Duality Numerics


(a)

Feasibility violations

(b)

False zeros

(c)

False nonzeros

(d)

Chambolle- Pock (CP) algorithm

## THANK YOU BORIS!!

Boris, thank you for your many mathematical gifts over the years!
and thank you for your warm friendship!

