

Paper notes

$$S = \{s_i : i=1 \dots\}$$

$$\text{conv}(S) \supseteq S$$

check if it
is zig-zagging

print gap
too

$$g_t^{FW} := \langle \nabla f(x^t), x^{(t)} - s_t \rangle$$

$$= \text{LO}(x^{(t)}) - \text{LO}(s_t)$$

$$= - \langle \nabla f(x^t), s_t - x^{(t)} \rangle$$

$$= - \langle \nabla f(x^t), x^t - u^t \rangle$$

$$= \text{LO}(u^t) - \text{LO}(x^t)$$

$$u^t \in \arg\max_{S^H} \langle \nabla f(x^t), u \rangle \parallel \text{closest to } x^t \text{ most far from } \text{grad.}$$

$$\uparrow \quad x^t - u^t \ll = \text{away from } u^t$$

$$\uparrow \uparrow \quad s^t - x^t \ll = \text{close to } s^t$$

\triangleright loc, convex or
strongly convex?

Regularization
from boosting VI
for h

approx ~~max~~ dist distribution
than ρ mix of
gaussian
 \downarrow
exponential?

improve opt $\&$
simplify entropy
regularization / norm
constraint

check if \times other
things on 1D & 2D
examples.

check distance
metrics for prob dist
using Optimal Transport

Estimate $L / \text{diam}(D)$

etc. on function
spaces for known f

check 1D run $\&$
behavior of line search,
fully conv. $\&$ fixed
(2D run too)

it was mentioned that
it stops at 1^+ iter.

see works of J-Lacoste
julien.

in FW we select

$$S = \{s_0, s_1, \dots\}$$

but $q_t \in \text{conv}(S)$

do something so that
 $\text{conv}(D) \supseteq D$ $\&$ is
more powerful.