

Convex Analysis

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1 Convex Functions

1.1 Subject

$$\text{epigraph} :: \prod V : \mathbb{R}\text{-VS} . \prod D \subset V . \left(D \rightarrow^{\infty} \mathbb{R} \right) \rightarrow ?(V \oplus \mathbb{R})$$

$$\text{epigraph}(f) = \text{epi } f := \{(x, \phi) | x \in D, \phi \in \mathbb{R}, \phi \geq f(x)\}$$

$$\text{Convex} :: \prod V : \mathbb{R}\text{-VS} . \prod D \subset V . ? \left(D \rightarrow^{\infty} \mathbb{R} \right)$$

$$f : \text{Convex} \iff \text{Convex}(V \oplus \mathbb{R}, \text{epi } f)$$

$$\text{effectiveDomain} :: \prod V : \mathbb{R}\text{-VS} . \prod D \subset V . \text{Convex}(V, D) \rightarrow ?D$$

$$\text{effectiveDomain}(f) = \text{dom } f := \pi_1 \text{epi } f$$

$$\text{DomainIsConvex} :: \forall V \in \mathbb{R}\text{-VS} . \forall D \subset V . \forall f : \text{Convex}(V, D) . \text{Convex}(V, \text{dom } f)$$

Proof =

As a linear image of convex set.

□

$$\text{ProperConvexFunction} :: \prod V : \mathbb{R}\text{-VS} . ? \text{Convex}(V, V) .$$

$$f : \text{ProperConvexFunction} \iff \forall x \in V . f(x) > -\infty \ \& \ \exists x \in V . f(x) < +\infty$$

InterpolationProperty ::

$$:: \forall V : \mathbb{R}\text{-VS} . \forall C : \text{Convex}(V) . \forall f : C \rightarrow (-\infty, +\infty] .$$

$$. \text{Convex}(V, C, f) \iff \forall x, y \in C . \forall \lambda \in [0, 1] .$$

$$. f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$$

Proof =

(\Rightarrow) : assume that f is convex.

Then f has convex epigraph.

Take arbitrary $x, y \in C$ and $\lambda \in [0, 1]$.

If f takes value $+\infty$ either in x or y , then the inequality follows, so assume the contrary.

Then $(x, f(x)), (y, f(y))$ trivially belong to the epigraph,

so by convexity $(\lambda x + (1 - \lambda)y, \lambda f(x) + (1 - \lambda)f(y))$ is also in epigraph.

The definition of epigraph produces the inequality $f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$.

(\Leftarrow) : now assume that inequality always hold.

Assume $(x, \phi), (y, \psi)$ belong to the epigraph and $\lambda \in [0, 1]$.

Then $\lambda \phi + (1 - \lambda)\psi \geq \lambda f(x) + (1 - \lambda)f(y) \geq f(\lambda x + (1 - \lambda)y)$.

So $\lambda(x, \phi) + (1 - \lambda)(y, \psi)$ belong to the epigraph.

Thus, epigraph is convex and f is convex.

JensensIneq ::

$:: \forall V : \mathbb{R}\text{-VS} . \forall C : \text{Convex}(V) . \forall f : C \rightarrow (-\infty, +\infty] .$

$. \forall n \in \mathbb{N} . \forall \lambda \in \mathbb{R}_+^n . \forall \mathbb{N} : \sum_{k=1}^n \lambda_k = 1 . \forall v \in V^n . f \left(\sum_{k=1}^n \lambda_k v_k \right) \leq \sum_{k=1}^n \lambda_k f(v_k)$

Proof =

Iterate the interpolation property.

□

SecondDerivativeConvexityTest :: $\forall I : \text{OpenInterval}(\mathbb{R}) . \forall f \in C^2(I) .$

$. \text{Convex}(\mathbb{R}, I, f) \iff f'' \geq 0$

Proof =

(\Rightarrow) : assume there is a $t \in I$ such that $f''(t) < 0$.

As f'' must be continous there is whole open interval (a, b) such that $f''(j) < 0$ for all $j \in (a, b)$.

Take some $x, y \in (a, b)$ with $x < y$ and define $z = \lambda x + (1 - \lambda)y$ for siome $\lambda \in (0, 1)$.

Then $f(z) - f(x) = \int_x^z f'(t) dt > f'(z)(z - x)$ and $f(y) - f(z) = \int_z^y f'(t) dt < f'(z)(y - z)$.

Then from definiton of z we get $f(z) > f(x) - (1 - \lambda)f'(z)(y - x)$ and $f(z) > f(y) + \lambda f'(z)(y - x)$.

By adding two inequalities with multipliers λ and $(1 - \lambda)$ one gets $f(\lambda x + (1 - \lambda)y) > \lambda f(x) + (1 - \lambda)f(y)$.

But this contradicts a convexity. .

(\Rightarrow) : use same inequalities but with different sign to prove the convexity.

□

ExponentIsConvexity :: $\forall \alpha \in \mathbb{R} . \text{Convex}(\mathbb{R}, \mathbb{R}, \Lambda t \in \mathbb{R} . e^{\alpha t})$

Proof =

write $f(t) = e^{\alpha t}$.

Then $f''(t) = \alpha^2 e^{\alpha t} \geq 0$.

□

MonomialConvexity1 :: $\forall p \in [1, +\infty) . \text{Convex}(\mathbb{R}, \mathbb{R}_{++}, \Lambda t \in \mathbb{R} . t^p)$

Proof =

Write $f(t) = t^p$.

Then $f''(t) = p(p - 1)t^{p-2} \geq 0$ for $t > 0$.

□

MonomialConvexity2 :: $\forall p \in [0, 1) . \text{Convex}(\mathbb{R}, \mathbb{R}_{++}, \Lambda t \in \mathbb{R} . -t^p)$

Proof =

Write $f(t) = t^p$.

Then $f''(t) = p(1 - p)t^{p-2} \geq 0$ for $t > 0$.

□

MonomialConvexity3 :: $\forall p \in (-\infty, 0] . \text{Convex}(\mathbb{R}, \mathbb{R}_{++}, \Lambda t \in \mathbb{R} . t^p)$

Proof =

write $f(t) = t^p$.

Then $f''(t) = p(p-1)t^{p-2} \geq 0$ for $t > 0$.

□

GeneralizedArcsinDerivativeIsConvex :: $\forall \alpha \in \mathbb{R}_{++} . \text{Convex}\left(\mathbb{R}, (-\alpha, \alpha), \Lambda t \in \mathbb{R} . \frac{1}{\sqrt{\alpha^2 - t^2}}\right)$

Proof =

Write $f(t) = \frac{1}{\sqrt{\alpha^2 - t^2}}$.

Then $f'(t) = \frac{t}{\sqrt{\alpha^2 - t^2}^3}$.

And $f''(t) = \frac{1}{\sqrt{\alpha^2 - t^2}^3} + \frac{3t^2}{\sqrt{\alpha^2 - t^2}^5} > 0$ for $t \in (-\alpha, \alpha)$.

□

NegativeLogIsConvex :: $\text{Convex}(\mathbb{R}, \mathbb{R}_{++}, \Lambda t \in \mathbb{R} . -\ln(t))$

Proof =

Write $f(t) = -\ln(t)$.

Then $f''(t) = \frac{1}{t^2} > 0$ for $t > 0$.

□

NegativeEntropyIsConvex :: $\text{Convex}(\mathbb{R}, \mathbb{R}_{++}, \Lambda t \in \mathbb{R} . t \ln(t))$

Proof =

Write $f(t) = t \ln(t)$.

Then $f'(t) = \ln(t) + 1$.

And $f''(t) = \frac{1}{t} > 0$ for $t > 0$.

□

Concave :: $\prod V : \mathbb{R}\text{-VS} . \prod D \subset V . ?(D \rightarrow \mathbb{R})$

$f : \text{Concave} \iff \text{Convex}(V, D, -f)$

SecondDerivativeConvexityTest2 :: $\forall V : \text{EuclideanSpace} . \forall U : \text{Open} \ \& \ \text{Convex}(V) . \forall f \in C^2(U) .$
 $\text{Convex}(\mathbb{R}, U, f) \iff \mathbf{D}^2 f \geq 0$

Proof =

For $x \in U$ and $v \in V \setminus \{0\}$ define $\phi_{x,v}(t) = f(x + tv)$ with a domain $I_{x,v} = \{t \in \mathbb{R} | x + tv \in C\}$.

Then f is convex iff every $\phi_{x,v}$ does.

But $\phi_{x,v}''(t) = \langle v, \mathbf{D}^2 f|_y v \rangle$, where $y = x + tv$.

So f is convex iff $\mathbf{D}^2 f$ is positive-semidefinite.

□

GeometricMeanIsConcave ::

$$:: \forall V : \text{EuclideanSpace} . \text{Concave} \left(V, V_{++}, \Lambda x \in V . \prod_{k=1}^n \sqrt[n]{x_k} \right) \quad \text{where } n = \dim V$$

Proof =

$$\text{write } f(x) = \prod_{k=1}^n \sqrt[n]{x_k}.$$

$$\text{Then } \nabla f|_x = \left(\frac{1}{n \sqrt[n]{x_i}^{n-1}} \prod_{j \neq i}^n \sqrt[n]{x_j} \right)_{i=1}^n.$$

$$\text{And } \mathbf{D}_{i,j}^2 f|_x = \frac{1}{n^2 \sqrt[n]{x_i x_j}^{n-1}} \prod_{k \neq i,j}^n \sqrt[n]{x_k} \text{ when } i \neq j, \text{ and } \mathbf{D}_{i,i}^2 f|_x = -\frac{n-1}{n^2 \sqrt[n]{x_i}^{2n-1}} \prod_{j \neq i}^n \sqrt[n]{x_j}.$$

$$\begin{aligned} \text{So, } \mathbf{D}^2 f|_x(v, v) &= -\frac{n-1}{n^2} \sum_{i=1}^n \frac{v_i^2}{\sqrt[n]{x_i}^{2n-1}} \prod_{j \neq i}^n \sqrt[n]{x_j} + \frac{1}{n^2} \sum_{i \neq j}^n \frac{v_i v_j}{\sqrt[n]{x_i x_j}^{n-1}} \prod_{k \neq i,j}^n \sqrt[n]{x_k} = \\ &= f(x) \left(-\frac{n-1}{n^2} \sum_{i=1}^n \frac{v_i^2}{x_i^2} + \frac{1}{n^2} \sum_{i \neq j}^n \frac{v_i v_j}{x_i x_j} \right) = -\frac{f(x)}{n^2} \left(n \sum_{i=1}^n \frac{v_i^2}{x_i^2} - \left(\sum_{i=1}^n \frac{v_i}{x_i} \right)^2 \right) \leq 0. \end{aligned}$$

This follows from obvious matching schema.

□

NormsAreConvex :: $\forall V : \mathbb{R}\text{-VS} . \forall \eta : \text{Norm}(V) \text{Convex}(V, V, \eta)$

Proof =

$$\text{Write } \eta(v) = \|v\|.$$

$$\text{Just use triangle inequality } \left\| \lambda x + (1-\lambda)y \right\| \leq \left\| \lambda x \right\| + \left\| (1-\lambda)y \right\| = \lambda \|x\| + (1-\lambda)\|y\|.$$

□

convexIndicator :: $\forall V : \mathbb{R}\text{-VS} . \text{Convex}(V) \rightarrow \text{Convex}(V, V)$

$$\text{convexIndicator}(C) = \Lambda x \in V . \chi(x|C) := \Lambda x \in V . \infty[x \in C^c]$$

supportFunction :: $\forall V : \mathbb{R}\text{-HIL} . \text{Convex}(V) \rightarrow \text{Convex}(V, V)$

$$\text{supportFunction}(C) = \Lambda x \in V . \chi^*(x|C) := \sup_{y \in C} \langle x, y \rangle$$

gauge :: $\forall V : \mathbb{R}\text{-VS} . \text{Convex}(V) \rightarrow \text{Convex}(V, V)$

$$\text{gauge}(C) = \Lambda x \in V . \gamma(x|C) := \Lambda x \in V . \inf \left\{ \lambda \in \mathbb{R}_{++} \mid x \in \lambda C \right\}$$

ConvexFunctionHasConvexLevelSets ::

$$:: \forall V \in \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) . \forall \alpha \in \mathbb{R}^{\infty} . \text{Convex} \left(V, \{v \in V : f(v) \geq \alpha\} \right)$$

Proof =

...

□

ConvexFunctionHasConvexStrictLevelSets ::

$$:: \forall V \in \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) . \forall \alpha \in \mathbb{R}^{\infty} . \text{Convex} \left(V, \{v \in V : f(v) > \alpha\} \right)$$

Proof =

...

ConvexlyBoundedRegionIsConvex ::

$$:: \forall V \in \mathbb{R}\text{-VS} . \forall I \in \text{SET} . \forall \alpha : I \rightarrow \mathbb{R}^{\infty} . \forall f : I \rightarrow \text{Convex}(V, V) . \text{Convex}\left(V, \{v \in V : \forall i \in I . f_i(v) > \alpha_i\}\right)$$

Proof =

$$\text{GeneralizedAMGMIneq} :: \forall n \in \mathbb{N} . \forall \lambda : \mathbb{R}_+^n . \forall x : \mathbb{R}_{++}^n . \forall \aleph : \sum_{i=1}^n \lambda_i = \sum_{i=1}^n \lambda_i x_i \geq \prod_{i=1}^n x_i^{\lambda_i}$$

Proof =

$$\text{By Jensen inequality for natural logarithm } \ln \left(\sum_{i=1}^n \lambda_i x_i \right) \geq \sum_{i=1}^n \lambda_i \ln(x_i).$$

$$\text{Then by exponentiating both parts } \sum_{i=1}^n \lambda_i x_i \geq \prod_{i=1}^n x_i^{\lambda_i}.$$

□

$$\text{PositivelyHomogeneous} :: \prod V : \mathbb{R}\text{-VS} . ?(V \rightarrow (-\infty, +\infty])$$

$$f : \text{PositivelyHomogeneous} \iff \forall v \in V . \forall \alpha \in \mathbb{R}_{++} . f(\alpha v) = \alpha f(v)$$

$$\text{PositiveHomogeneousZeroPositivity} :: \forall V : \mathbb{R}\text{-VS} . \forall f : \text{PositivelyHomogeneous}(V) . f(0) \geq 0$$

Proof =

$$\text{Note that } f(0) = f(t0) = tf(0) \text{ for all } t \in \mathbb{R}_{++}.$$

$$\text{This means that } f(0) \text{ is either } 0 \text{ or } +\infty.$$

□

$$\text{PositiveHomogeneousConvexity} :: \forall V : \mathbb{R}\text{-VS} . \forall f : \text{PositivelyHomogeneous}(V) .$$

$$. \text{Convex}(V, V, f) \iff \forall x, y \in V . f(x + y) \leq f(x) + f(y)$$

Proof =

$$(\Rightarrow) : \text{assume } f \text{ is convex.}$$

$$\text{Then } f(x + y) = f\left(\frac{2}{2}x + \frac{2}{2}y\right) \leq \frac{1}{2}f(2x) + \frac{1}{2}f(2y) = f(x) + f(y) \text{ for any } x, y \in V.$$

$$(\Leftarrow) : \text{assume the inequality holds .}$$

$$\text{Then } f(\lambda x + (1 - \lambda)y) \leq f(\lambda x) + f((1 - \lambda)y) = \lambda f(x) + (1 - \lambda)f(y) \text{ when } \lambda \in (0, 1) \text{ and } x, y \in V .$$

$$\text{Otherwise, when } \lambda = 0, 1, \text{ convexity condition holds trivially.}$$

□

$$\text{Conic} := \lambda V \in \mathbb{R}\text{-VS} . \text{Convex}(V, V) \times \text{PositivelyHomogeneous}(V) : \mathbb{R}\text{-VS} \rightarrow \text{Type};$$

$$\text{ConicIneq} :: \forall V : \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) \ \& \ \text{PositivelyHomogeneous}(V) . \forall n \in \mathbb{N} . \forall x \in V^n .$$

$$. \forall \lambda \in \mathbb{R}_{++}^n . f\left(\sum_{i=1}^n \lambda_i x_i\right) \leq \sum_{i=1}^n \lambda_i f(x_i)$$

Proof =

$$\text{Iterate previous theorem.}$$

□

ConicEpigraph :: $\forall V \in \mathbb{R}\text{-VS} . \forall f : V \rightarrow (-\infty, +\infty) . \text{Conic}(V, f) \iff \text{ConvexCone}(V, \text{epi } f)$

Proof =

...

□

ConicIsSupersymmetric :: $\forall V \in \mathbb{R}\text{-VS} . \forall f \in \text{Conic}(f) . \forall v \in V . f(v) \geq -f(-v)$

Proof =

Write $f(x) + f(-x) \geq f(x - x) = f(x) \geq 0$.

So $f(x) \geq -f(-x)$.

□

ConicIsLinearIffSymmetric :: $\forall V \in \mathbb{R}\text{-VS} . \forall f \in \text{Conic}(f) . f \in V^* \iff \forall v \in V . f(-v) = -f(v)$

Proof =

(\Rightarrow) : this is trivial.

(\Leftarrow) : assume that the property holds .

Let $x, y \in V$.

Then $f(x) + f(y) \geq f(x + y) \geq -f(-x - y) \geq -f(-x) - f(-y) = f(x) + f(y)$.

This mean $f(x) + f(y) = f(x + y)$.

But as x and y were arbitrary f must be additive and hence linear.

□

StrictlyConvexFunction :: $\prod V : \mathbb{R}\text{-VS} . ?\text{ProperConvexFunction}(V)$

$f : \text{StrictlyConvexFunction} \iff$

$$\iff \forall \lambda \in [0, 1] . \forall x, y \in V . \forall \lambda : x \neq y . f(\lambda x + (1 - \lambda)y) < \lambda f(x) + (1 - \lambda)f(y)$$

SquareNoremIsStrictlyConvex :: $\forall V : \text{NormedSpace}(\mathbb{R}) . \text{StrictConvexFunction}(V, \|\bullet\|^2)$

Proof =

$$\forall x, y \in V, \lambda \in (0, 1) . \left\| \lambda x + (1 - \lambda)y \right\|^2 \leq \left(\lambda \|x\| + (1 - \lambda)\|y\| \right)^2 < \lambda^2 \|x\|^2 + (1 - \lambda)^2 \|y\|^2 < \lambda \|x\|^2 + (1 - \lambda)\|y\|^2 .$$

□

1.2 Convexity Preserving Operations

ConvexComposition :: $\forall V \in \mathbb{R}\text{-VS} . \forall D \subset V \forall f : \text{Convex}(V, D) . \forall \phi : \text{Convex} \ \& \ \text{Increasing}(\mathbb{R}, \mathbb{R}) . \text{Convex}(V,$

Proof =

Assume $x, y \in \text{dom } f, \lambda \in [0, 1]$.

Then $\phi\left(f(\lambda x + (1 - \lambda)y)\right) \leq \phi\left(\lambda f(x) + (1 - \lambda)f(y)\right) \leq \lambda \phi \circ f(x) + (1 - \lambda)\phi \circ f(y)$.

□

ConvexFunctionFromSet :: $\forall V \in \mathbb{R}\text{-VS} . \forall C : \text{Convex}(V \oplus \mathbb{R}) . \text{Convex}\left(V, V, \Lambda v \in V . \inf \{t \mid (v, t) \in C\}\right)$

Proof =

This is function has cinvex epigraph.

□

InfimalConvolutionIsConvex :: $\forall V \in \mathbb{R}\text{-VS} . \forall n \in \mathbb{N} . \forall f : \{1, \dots, n\} \rightarrow \text{ProperConvexFunction}(V) .$

$. \text{Convex}\left(V, V, \Lambda x \in V . \inf \left\{ \sum_{k=1}^n f_k(v_k) \mid v \in V^n, \sum_{k=1}^n v_k = x \right\}\right)$

Proof =

Let $g = \inf \left\{ \sum_{k=1}^n f_k(v_k) \mid v \in V^n, \sum_{k=1}^n v_k = x \right\}$.

$C = \sum_{k=1}^n \text{epi } f_k$ is convex.

A tuple $(x, \phi) \in C$ if there is a sequence $(v, \psi) \in (V \oplus \mathbb{R})^n$ such that $x = \sum_{k=1}^n v_k, \phi = \sum_{k=1}^n \psi_k$

and $f(v_k) \leq \psi_k$ for every $k \in \{1, \dots, n\}$.

Thus $\phi = \sum_{k=1}^n \psi_k \geq \sum_{k=1}^n f(v_k) \geq g(x)$, so $(x, \phi) \in \text{epi } g$.

Then g can be constructed from set C .

□

infimalConvolution :: $\prod_{V \in \mathbb{R}\text{-VS}} \prod_{n=1}^{\infty} \left(\{1, \dots, n\} \rightarrow \text{ProperConvexFunction}(V) \right) \rightarrow \text{Convex}(V, V)$

infimalConvolution $(f) = \bigsqcup_{k=1}^n f_i := \inf \left\{ \sum_{k=1}^n f_k(v_k) \mid v \in V^n, \sum_{k=1}^n v_k = x \right\}$

ConvexDelta :: $\prod_{V \in \mathbb{R}\text{-VS}} V \rightarrow \text{ProperConvexFunction}(V)$

convexDelta $(a) = \delta_a := \Lambda x \in V . \text{if } x = a \text{ then } 0 \text{ else } +\infty$

GraphTranslationByInfimalConvolution :: $\forall V \in \mathbb{R}\text{-VS} . \forall f : \text{ProperConvexFunction}(V) . \forall a, v \in V . (\delta_a \square f)(v) = \min \{ f(v - a), +\infty \}$

Proof =

Clearly $(\delta_a \square f)(v) = \min \{ f(v - a), +\infty \}$.

□

InfimalConvolutionDomain :: $\forall V \in \mathbb{R}\text{-VS} . \forall f : \text{ProperConvexFunction}(V) . \forall a, v \in V . \text{dom}(f \square g) = \text{dom } f$

Proof =

Obvious.

□

DistanceExpression :: $\forall V : \text{NormedSpace}(\mathbb{R}) . \forall C : \text{Convex}(V) . d_V(C, \bullet) = \| \bullet \| \square \chi(\bullet | C)$

Proof =

□

InfimalConvolutionDefinesCommutativeMonoid ::

$:: \forall V \in \mathbb{R}\text{-VS} .$

$. \text{CommutativeMonoid}(\text{Convex}(V, V), \Lambda f, g \in \text{Convex}(V, V) . \Lambda x \in V . \inf \{ \phi \mid (v, \phi) \in (\text{epi } f + \text{epi } g) \})$

Proof =

δ_0 is a neutral element, comutativity and associativity is almost obvious.

□

rightScalarMultiplication :: $\prod_{V \in \mathbb{R}\text{-VS}} . \text{Convex}(V, V) \rightarrow \mathbb{R}_+ \rightarrow \text{Convex}(V, V)$

rightScalarMultiplication $(f, \lambda) = f\lambda := \text{ConvexFunctionFromSet}(V, \lambda \text{epi } f)$

RightScalarMultiplicationExpression ::

$:: \forall V \in \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) . \forall \lambda \in \mathbb{R}_{++} . \forall x \in X . f\lambda(x) = \lambda f(\lambda^{-1}x)$

Proof =

Obvious.

□

RightScalarMultiplicationByZero :: $\forall V \in \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) . f0(x) = \delta(0)$

Proof =

Obvious.

□

ConicityByRightMultiplication :: $\forall V \in \mathbb{R}\text{-VS} . \forall f : \text{Convex}(V, V) . \text{Conic}(V, f) \iff \forall \lambda \in \mathbb{R}_{++} . f\lambda = f$

Proof =

Follows from the expression for right multiplication.

□

$$\text{conicClosure} :: \prod_{V \in \mathbb{R}\text{-VS}} \text{ConvexFunction}(V) \rightarrow \text{Conic}(V)$$

$$\text{conicClosure}(\text{cone } f) := \text{ConvexFunctionFromSet}\left(V, \text{cone epi } f\right)$$

$$\text{GaugeExpression} :: \forall V \in \mathbb{R}\text{-VS} . \forall C : \text{Convex} \ \& \ \text{NonEmpty}(V) . \gamma(\bullet|C) = \text{cone}\left(\chi(\bullet|C) + 1\right)$$

Proof =

$$(x, \phi) \in \text{epi } \gamma(\bullet|C) \text{ iff } x \in \lambda C \text{ and } 0 < \lambda \leq \phi.$$

$$\text{This means that } (x, \lambda) \in \text{cone } C \times \{1\} \subset \text{cone epi}\left(\chi(\bullet|C) + 1\right).$$

$$\text{So } (x, \phi) \in \text{cone } C \times \{\phi/\lambda\} \subset \text{cone epi}\left(\chi(\bullet|C) + 1\right) = \text{epi cone}\left(\chi(\bullet|C) + 1\right).$$

$$\text{On the other hand id } (x, \psi) \in \text{epi cone}\left(\chi(\bullet|C) + 1\right) \text{ then there exists } \lambda \in \mathbb{R}_{++} \text{ such that } \lambda x \in C \text{ and } \lambda \psi \geq 1 .$$

$$\text{But this means that } \psi \geq \lambda^{-1} \geq \gamma(x|C).$$

$$\text{Thus } (x, \psi) \in \gamma(\bullet|C) .$$

And both functions are equal by equality of epigraphs.

□

$$\text{SupremumIsConvex} :: \forall V \in \mathbb{R}\text{-VS} . \forall I \in \text{SET} . \forall f : I \rightarrow \text{ConvexFunction}(V) . \text{ConvexFunction}(V, \sup_{i \in I} f_i)$$

Proof =

$$\text{epi } \sup_{i \in I} f_i = \bigcap_{i \in I} \text{epi } f_i \text{ is convex.}$$

$$\text{convexHull} :: \prod_{V \in \mathbb{R}\text{-VS}} \prod_{I \in \text{SET}} \left(I \rightarrow V \rightarrow^{\infty} \mathbb{R} \right) \rightarrow \text{ConvexFunction}(V)$$

$$\text{convexHull}(f) = \text{conv}_{i \in I} f_i := \text{ConvexFunctionFromSet}\left(V, \text{conv} \bigcup_{i \in I} \text{epi } f_i\right)$$

$$\text{ConvexHullExpression} :: \forall V \in \mathbb{R}\text{-VS} . \forall I \in \text{SET} . \forall f : I \rightarrow V \rightarrow (-\infty, +\infty] . \forall v \in V .$$

$$. \text{conv}_{i \in I} f_i(x) = \inf \left\{ \sum_{i \in I} \lambda_i f_i(v_i) \middle| \lambda \in \mathbb{R}_+^{\oplus I}, v : I \rightarrow V, \sum_{i \in I} \lambda_i = 1, \sum_{i \in I} \lambda_i v_i = x \right\}$$

Proof =

This follows from the thorough examination of the definition.

$$\text{convexPullback} :: \prod V, W \in \mathbb{R}\text{-VS} . \mathbb{R}\text{-VS}(V, W) \rightarrow \text{ConvexFunction}(W) \rightarrow \text{ConvexFunction}(V)$$

$$\text{convexPullback}(f, T) = fT := f \circ T$$

$$\text{convexPushforward} :: \prod V, W \in \mathbb{R}\text{-VS} . \mathbb{R}\text{-VS}(V, W) \rightarrow \text{ConvexFunction}(V) \rightarrow \text{ConvexFunction}(W)$$

$$\text{convexPullback}(f, T) = T_* f := \Lambda w \in W . \inf\{f(v) | w = Tv\}$$

1.3 Metric and Topological Properties

SphericalBound ::

$:: \forall V \in \text{BAN}(\mathbb{R}) . \forall f : \text{ProperConvexFunction}(V) . \forall c \in \text{dom } f . \forall \rho \in \mathbb{R}_{++} . \forall \eta : \eta < +\infty . \forall \alpha \in (0, 1) .$
 $. \forall x \in \mathbb{B}_V(c, \alpha\rho) . \left| f(x) - f(c) \right| \leq \alpha \left(\eta - f(c) \right)$
where $\eta = \sup f \left(\mathbb{B}_V(c, \rho) \right)$

Proof =

By convexity $f(x) - f(c) = f \left((1 - \alpha)c + \alpha \left(\frac{x - (1 - \alpha)c}{\alpha} \right) \right) - f(c) \leq \alpha \left(f \left(c + \frac{x - c}{\alpha} \right) - f(c) \right)$.

If $x \in \mathbb{B}_V(c, \alpha\rho)$ then $c + \frac{x - c}{\alpha} \in \mathbb{B}_V(c, \rho)$.

So $f(x) - f(c) \leq \alpha \left(\eta - f(c) \right)$.

On the other hand $f(c) - f(x) = f \left(\frac{x}{1 + \alpha} + \frac{\alpha}{1 - \alpha} \frac{(1 + \alpha)c - x}{\alpha} \right) - f(x) \leq$
 $\leq \frac{\alpha}{1 + \alpha} \left(f \left(c + \frac{c - x}{\alpha} \right) - f(x) \right) \leq \frac{\alpha}{1 + \alpha} (\eta - f(x)) = \frac{\alpha}{1 + \alpha} (\eta - f(c)) + \frac{\alpha}{1 + \alpha} (f(c) - f(x))$.

So by rearranging inequalities one gets $f(c) - f(x) \leq \alpha (\eta - f(x))$.

LocalLipschitzContinuity ::

$:: \forall V \in \text{BAN}(\mathbb{R}) . \forall f : \text{ProperConvexFunction}(V) . \forall c \in \text{dom } f . \forall \rho \in \mathbb{R}_{++} . \forall \delta : \delta < +\infty .$
 $. \left(\frac{\delta}{\rho} \right) \text{-Lip} \left(\mathbb{B}(c, \rho), \mathbb{R}, f|_{\mathbb{B}(c, \rho)} \right)$
where $\eta = \text{diam } f \left(\mathbb{B}_V(c, \rho) \right)$

Proof =

Assume $x, y \in \mathbb{B}_V(c, \rho)$ such that $x \neq y$.

Let $\alpha = \frac{\|x - y\|}{\|x - y\| + \rho} < \frac{\|x - y\|}{\rho}$ and $z = x + \frac{1 - \alpha}{\alpha}(x - y)$.

Then, $\|z - c\| \leq \|z - x\| + \|x - c\| \leq \frac{1 - \alpha}{\alpha} \|x - y\| + \rho \leq 2\rho$, so $z \in \mathbb{B}_V(c, 2\rho)$.

Thus, by convexity $f(x) = f(\alpha z + (1 - \alpha)y) \leq f(y) + \alpha(f(z) - f(y)) \leq f(y) + \alpha\delta \leq f(y) + \frac{\delta}{\rho} \|x - y\|$.

From symmetry $|f(x) - f(y)| \leq \frac{\delta}{\rho} \|x - y\|$.

□

ContinuityByBound ::

$$\begin{aligned} &:: \forall V \in \mathbf{BAN}(\mathbb{R}) . \forall f : \mathbf{ProperConvexFunction}(V) . \forall x \in V . \forall U \in \mathcal{U}(x) . \\ & . \forall \mathbb{N} : \mathbf{Bounded}(V, U, f|_U) . \text{int dom } f \xrightarrow{f|_{\text{int dom } f}} \mathbb{R} : \mathbf{TOP} \end{aligned}$$

Proof =

Assume $v \in U \cap \text{rel int dom } f$.

As v belongs to relative interior there are $w \in \text{dom } f, \rho \in \mathbb{R}_{++}, \lambda \in (0, 1)$

such that $v \in \lambda \mathbb{B}(x, \rho) + (1 - \lambda)w$ and $\mathbb{B}(x, \rho) \subset U$.

Then $f(x) \leq \lambda f(y) + (1 - \lambda)f(w) \leq \lambda \beta + (1 - \lambda)f(w)$,

where $x \in \lambda \mathbb{B}(x, \rho) + (1 - \lambda)w, y \in \mathbb{B}(y, \rho)$ and β is the bound for U .

So by the previous theorem f is locally Lipschitz and continuous on $\text{rel int dom } f$ and so f is actually continuous on $\text{rel int dom } f$.

□

ContinuityByFiniteDimension ::

$$:: \forall V \in \mathbf{EuclideanSpace} . \forall f : \mathbf{ProperConvexFunction}(V) . \text{int dom } f \xrightarrow{f|_{\text{int dom } f}} \mathbb{R} : \mathbf{TOP}$$

Proof =

Let $n = \dim V$.

Assume $x \in V$ and $\rho \in \mathbb{R}_{++}$ such that $\mathbb{B}(x, \rho) \subset \text{dom } f$.

Then there exists a simplicital set $\{v_1, \dots, v_{n+1}\}$ such that $\mathbb{B}(x, \rho) \subset \text{conv}(v_1, \dots, v_{n+1}) \subset \text{dom } f$.

But this means that $\sup f(\mathbb{B}(x, \rho)) \leq \max_{i=1, \dots, n+1} f(v_i)$.

So by the previous theorem f is locally Lipschitz and continuous on $\text{rel int dom } f$ and so f is actually continuous on $\text{rel int dom } f$.

□

NonEmptyInteriorCondition :: $\forall V \in \mathbf{BAN}(\mathbb{R}) . \forall f : \mathbf{ProperConvexFunction}(V) . \forall x \in V . \forall U \in \mathcal{U}(x) .$
 $. \forall \mathbb{N} : \mathbf{Bounded}(V, U, f|_U) . \text{int epi } f \neq \emptyset$

Proof =

We can find radius ρ Lipschitz constant β such that $|f(v) - f(w)| \leq \beta \|v - w\| \leq \beta \rho$ for every $v, w \in \mathbb{B}(x, \rho)$.

Select $\delta \in (2\beta\rho, +\infty)$ and set $\gamma = \min(\rho, \delta/2) > 0$.

And let $(y, \phi) \in V \oplus \mathbb{R}$ such that $\|(y, \phi) - (x, f(x) + \rho)\|^2 \leq \gamma^2$.

Then $\|y - x\| \leq \gamma \leq \rho$ and $|\phi - (f(x) + \delta)| \leq \gamma \leq \delta/2$.

So, $f(y) < f(x) + \delta/2 = f(x) + \delta - \delta/2 \leq f(x) + \delta - \gamma \leq \phi$.

Thus $(y, \phi) \in \text{epi } f$.

As (y, ϕ) was arbitrary, the whole sphere is a subset of $\text{epi } f$.

So $\text{int epi } f \neq \emptyset$.

□

InteriorLevelSet ::

$:: \forall V : \mathbb{R}\text{-BAN} . \forall f : \text{LowerSemicontinuous} \left(V, (-\infty, +\infty] \right) . \forall x \in V . \forall \mathbb{N} : f(x) < 0 . x \in \text{int } f^{-1}(-\infty, 0]$

Proof =

...

□

ConvexLevelSetInterior :: $\forall V : \mathbb{R}\text{-BAN} . \forall f : \text{ConvexFunction}(V) . \forall x \in V . \forall \mathbb{N} : f(x) < 0 .$
 $. \text{int } f^{-1}(-\infty, 0] \subset f^{-1}(-\infty, 0)$

Proof =

...

□

ConvexUsCLevelSetInteriorEq ::

$. \forall V : \mathbb{R}\text{-BAN} . \forall f : \text{ConvexFunction}(V) . \forall \mathbb{N} : \text{LowerSemicontinuous} \left(f|_{f^{-1}(-\infty, 0)}, (-\infty, 0) \right) .$
 $. \text{int } f^{-1}(-\infty, 0] = \text{int } f^{-1}(-\infty, 0)$

Proof =

...

□

ConvexEucLevelSetInteriorEq ::

$. \forall V \in \text{EuclideanSpace} . \forall f : \text{ConvexFunction}(V) . \forall \mathbb{N} : \text{dom } f \in \mathcal{T}(V) . \text{int } f^{-1}(-\infty, 0] = \text{int } f^{-1}(-\infty, 0)$

Proof =

...

□

1.4 Closures

LowerSemicontinuityByEpigraph :: $\forall V \in \text{EuclideanSpace} . \forall f : V \rightarrow \mathbb{R} \text{-TOPVS} .$

LowerSemicontinuous $(V, \mathbb{R}, f) \iff \text{Closed}(V \oplus \mathbb{R}, \text{epi } f)$

Proof =

If f is lower semicontinuous then $\liminf_{x \rightarrow v} f(x) = f(v)$.

So the epigraph must be closed.

Thus, result is basically obvious.

closure :: $\prod_{V \in \mathbb{R}\text{-TOPVS}} \text{ConvexFunction}(V) \rightarrow (\text{ConvexFunction}(V) \ \& \ \text{LowerSemicontinuous}(V, \mathbb{R}))$

closure $(f) = \text{cl } f := \text{if } f > -\infty \text{ then } \text{FunctionFromSet}(V, \text{cl epi } f) \text{ else } -\infty$

ClosedFunction :: $\prod V \in \mathbb{R}\text{-TOPVS} . ?\text{ConvexFunction}(V)$

$f : \text{ClosedFunction} \iff \text{cl } f = f$

ImproperDomain :: $\forall V \in \mathbb{R}\text{-TOPVS} . \forall f : \text{ConvexFunction}(V) .$

$. \forall \mathbb{N} : -\infty \in \text{Im } f . \forall x \in \text{rel int dom } f . f(x) = -\infty$

Proof =

If there is a point $p \in V$ such that $f(p) = -\infty$ then $p \in \text{dom } f$.

Also assume that $u \in \text{rel int dom } f$.

By properties of relative interior there exists $x \in \text{dom } f$ such that $u \in (p, x)$.

So there is $\lambda \in (0, 1)$ such that $u = \lambda u + (1 - \lambda)x$.

But $(p, \alpha) \in \text{epi } f$ for any arbitrary $\alpha \in \mathbb{R}$.

Thus $(u, \lambda\alpha + (1 - \lambda)f(x)) \in \text{epi } f$ for any α .

And by taking the limit $\alpha \rightarrow -\infty$ we see that it must be the case that $f(u) = -\infty$.

□

ContinuityByClosedness ::

$:: \forall V \in \text{BAN}(\mathbb{R}) . \forall f : \text{ClosedFunction}(V) . \text{int dom } f \xrightarrow{f|_{\text{int dom } f}} \mathbb{R} : \text{TOP}$

Proof =

...

□

ConvexLsCLevelSetInteriorEq ::

$. \forall V \in \text{EuclideanSpace} . \forall f : \text{ClosedFunction}(V) . \forall \mathbb{N} : \text{dom } f \in \mathcal{T}(V) .$

$. \text{int } f^{-1}(-\infty, 0] = \text{int } f^{-1}(-\infty, 0)$

Proof =

...

□

ClosedFunctionSupremum :: $\forall V \in \mathbb{R}\text{-TOPVS} . \forall I \in \text{SET} . \forall f : I \rightarrow \text{ClosedFunction}(V) .$
 $. \sup f \in \text{ClosedFunction}(V)$

Proof =

The epigraph of the supremum is the intersection of epigraphs.

Then use the fact that intersection of closed sets is closed.

□

ClosedFunctionSum ::

$. \forall V \in \mathbb{R}\text{-TOPVS} . \forall f, g \in \text{ClosedFunction} \ \& \ \text{ProperConvexFunction}(V) .$

$. f + g \in \text{ClosedFunction} \ \& \ \text{ProperConvexFunction}(V)$

Proof =

...

□

ClosedFunctionSum2 ::

$. \forall V \in \mathbb{R}\text{-TOPVS} . \forall I \in \text{SET} . \forall f : \mathbb{N} \rightarrow \text{ClosedFunction}(V) .$

$. \forall \mathbb{N} : \inf_{i \in I} f_i \geq 0 . \sum_{i \in I} f_i \in \text{ClosedFunction}(V)$

Proof =

...

□

DegenerateClosedForm ::

$. \forall V \in \mathbb{R}\text{-TOPVS} . \forall f : \text{ClosedFunction}(V) . \forall \mathbb{N} : -\infty \in \text{Im } f . \text{Im } f = \{-\infty, +\infty\}$

Proof =

...

□

ClosureLevelSets ::

$. \forall V \in \mathbb{R}\text{-TOPVS} . \forall f : \text{ConvexFunction}(V) . \text{cl} \left(f^{-1}(-\infty, 0) \right) = \text{cl} \left(f^{-1}(-\infty, 0] \right) = \left(\text{cl } f \right)^{-1}(-\infty, 0)$

Proof =

...

□

ProperClosedFunction := **ProperConvexFunction** & **ClosedFunction** : $\mathbb{R}\text{-TOPVS} \rightarrow \text{Type}$;

ConvexLimit :: $\forall V \in \mathbb{R}\text{-TOPVS} . \forall f : \text{ProperClosedFunction}(V) . \forall x \in \text{dom } f . \forall y \in V . \lim_{\lambda \downarrow 0} g(\lambda) = f(y)$

where $g = \lambda \mapsto f(\lambda x + (1 - \lambda)y)$

Proof =

In $f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$ the majorant is continuous in *lambda*.

So in $f(x) \leq \lim_{\lambda \rightarrow \infty} f(\lambda x + (1 - \lambda)y) \leq f(x)$ holds.

□

ProperConvexContinuity :: $f \in \text{ProperClosedFunction}(\mathbb{R}) \cdot f|_{\text{cl dom } f} \in \text{TOP}\left(\text{cl dom } f, \overset{\infty}{\mathbb{R}}\right)$

Proof =

Use convex limits as above.

□

ConvexExtension ::

$\forall V \in \mathbb{R}\text{-TOPVS} \cdot \forall f, g \in \text{ProperClosedFunction}(V \oplus \mathbb{R}) \cdot$

$\cdot \forall \aleph : \text{dom } f \cup \text{dom } g \subset V \oplus \mathbb{R}_+ \cdot \forall \sqsupset : f|_{V \times \mathbb{R}_{++}} = g|_{V \times \mathbb{R}_{++}} \cdot f = g$

Proof =

...

□

1.5 Affine Minorization

1.6 Recession

2 Duality

3 (Sub)differential Calculus

4 From Optimization to Convex Algebra

Sources

1. Convex Analysis — R. T. Rockaffeler 1972
2. Convex Analysis and Monotone Operator Theory in Hilbert Spaces – H. H. Bauschke, P. L. Combettes 2010