

Nonparametric estimation under Shape Restrictions



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Outline: Five Lectures on Shape Restrictions

- L1: Monotone functions: maximum likelihood and least squares
- L2: Optimality of the MLE of a monotone density (and comparisons?)
- L3: Estimation of convex and k -monotone density functions
- L4: Estimation of log-concave densities:
 $d = 1$ and beyond
- **L5: More on higher dimensions ...
... and some open problems**

Outline: Lecture 5

- A: Some multivariate shape-constrained classes
 - ▷ Log-concave & h -convex on \mathbb{R}^d
 - ▷ Block-decreasing on \mathbb{R}_+^d
 - ▷ Scale mixtures of uniform on \mathbb{R}_+^d
 - ▷ h -convex on \mathbb{R}_+^d with h increasing.
- B: Review of available theory; MLEs for multivariate classes.
- C: Some alternative classes in \mathbb{R} :
Bondesson's hyperbolically monotone classes.
- D: Open problems and questions: \mathbb{R}^1
- E: Some problems and questions: \mathbb{R}^d

A. Some multivariate shape-constrained classes

- “Block decreasing” densities on $R^{+d} = [0, \infty)^d$
- Monotone with non-negative increments on rectangles (as for a multivariate d.f.)
- Convex and decreasing
- k –monotone; completely monotone
- log–concave; s –concave; h – transform of convex (or concave)

A. Some multivariate shape-constrained classes

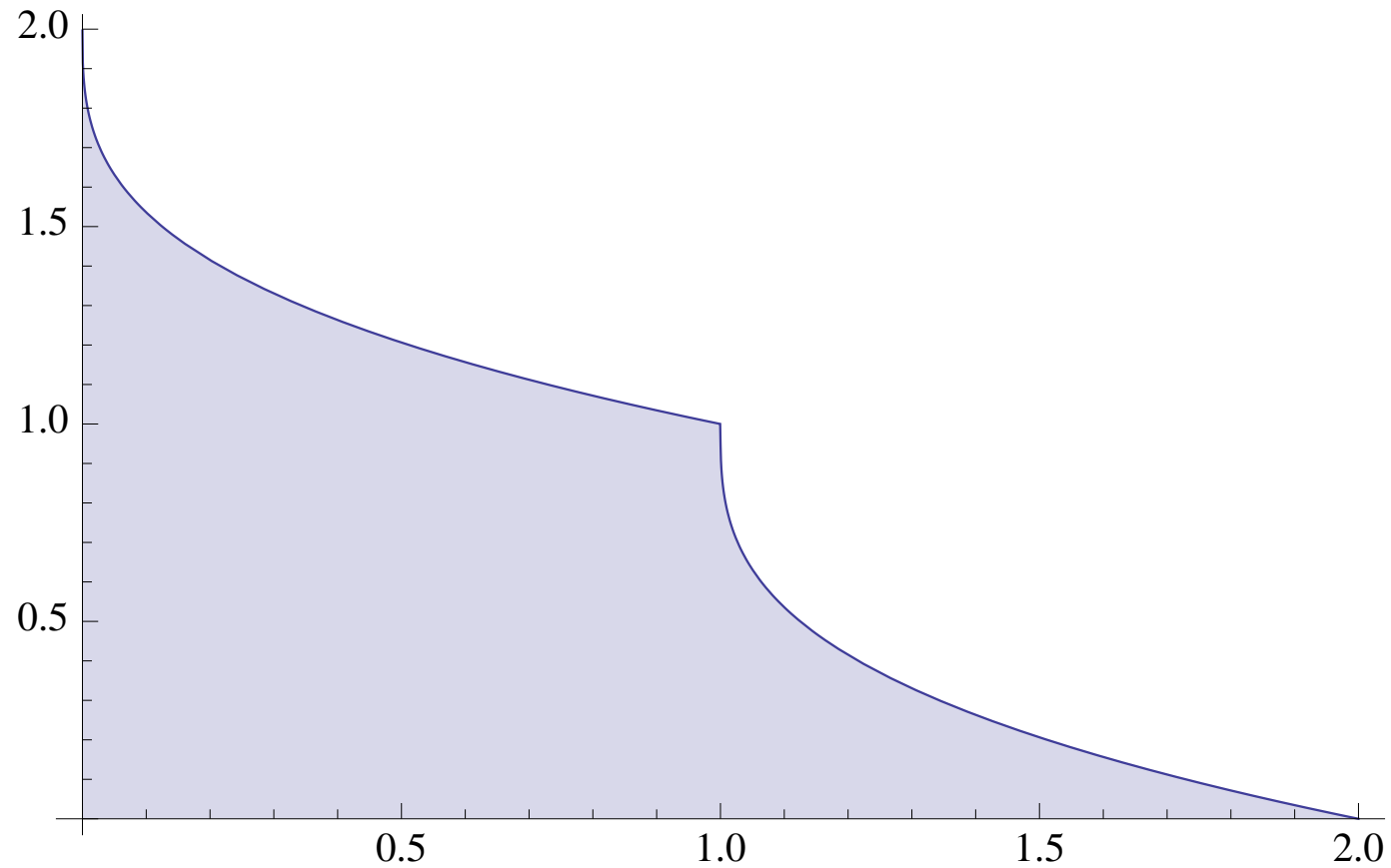
Block-decreasing densities on $\mathbb{R}^{+d} = [0, \infty)^d$

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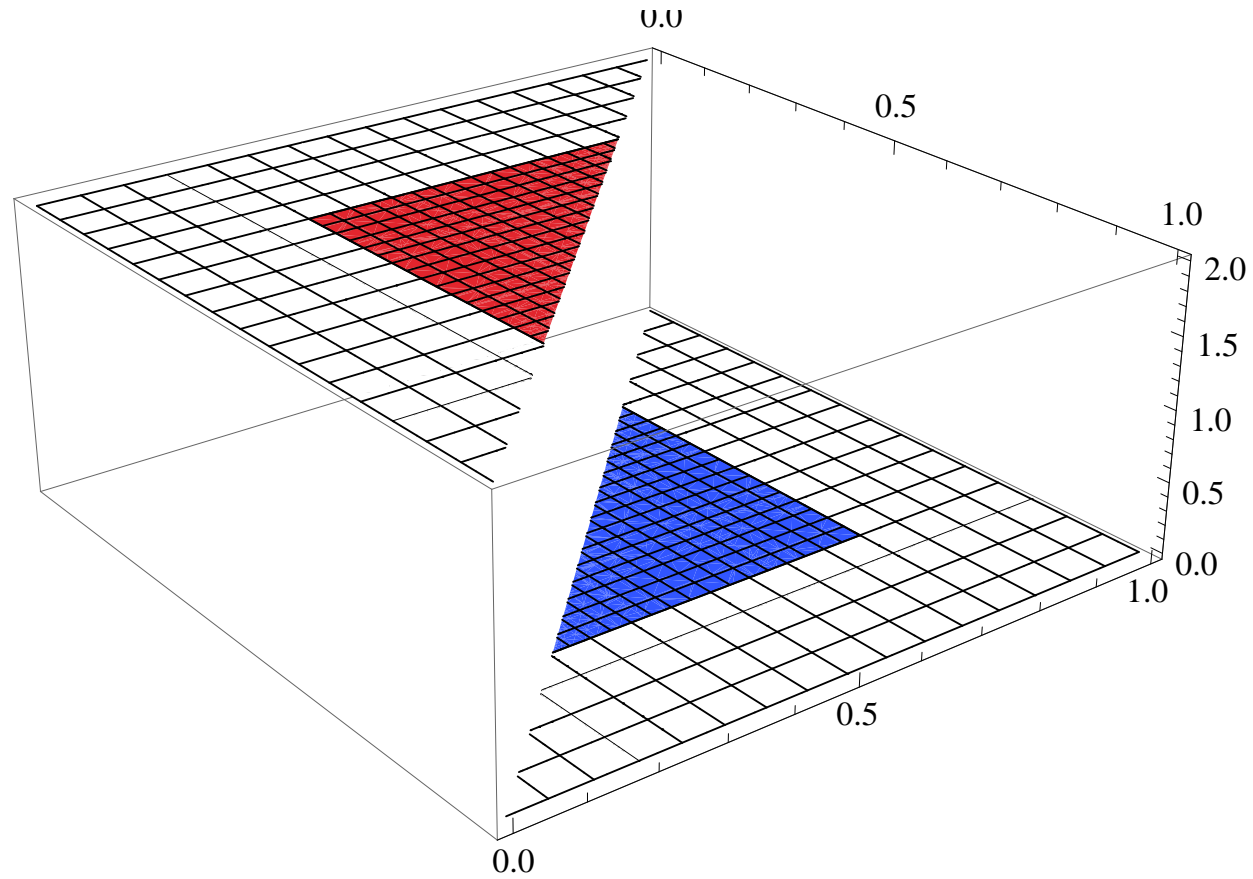
$$\mathcal{BD}(\mathbb{R}^d) = \left\{ \begin{array}{l} f : [0, \infty)^d \rightarrow \mathbb{R}^+ \mid \int f(\underline{x}) d\underline{x} = 1, \\ f(\underline{x} + h\underline{e}_j) \leq f(\underline{x}) \\ \text{for all basis vectors } \underline{e}_j, j = 1, \dots, d, h > 0 \end{array} \right\}.$$

- Polonik (1995, 1998): MLE exists and coincides with “silhouette estimator”.
- Biau and Devroye (2003): global minimax lower bounds ... and showed that a generalization of Birgé’s histogram estimator achieves the bounds.
- Pavlides (2008, 2009): asymptotic minimax lower bounds for estimation of $f(x_0)$

A. Some multivariate shape-constrained classes



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A. Some multivariate shape-constrained classes

**Monotone with non-negative increments on rectangles
(as for a multivariate d.f.)**

= “Scale mixtures of uniform densities” on \mathbb{R}^{+d}

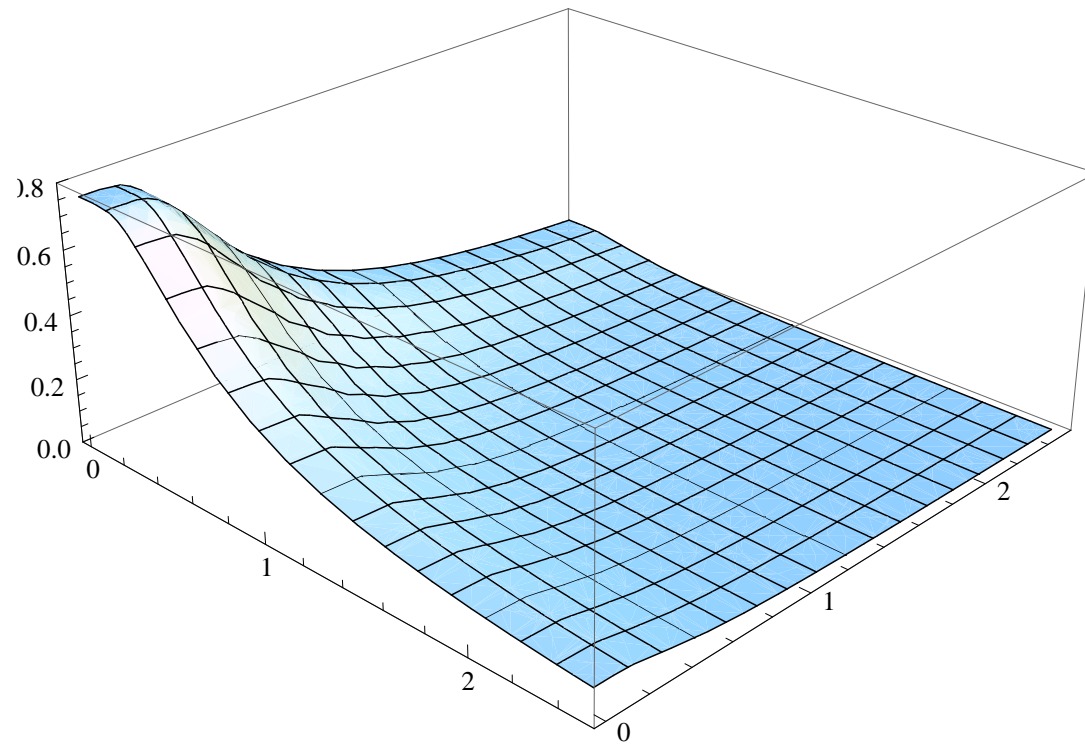
$$f(\underline{x}) = \int_{\mathbb{R}^{+d}} \frac{1}{\prod_{j=1}^d y_j} \mathbf{1}_{[0, \underline{y}]}(\underline{x}) dG(\underline{y})$$

for some probability distribution G on \mathbb{R}^{+d} .

Example: $dG(y_1, y_2) = (y_1 y_2)^{-2} g(1/y_1, 1/y_2, \theta) dy_1 dy_2$ with

$$g(u, v, \theta) = \{(1 + \theta u)(1 + \theta v) - \theta\} \exp(-u - v - \theta uv), \quad \theta = .4,$$

A. Some multivariate shape-constrained classes



A. Some multivariate shape-constrained classes

Convex and decreasing on \mathbb{R}^d

- Seregin (2010)'s increasing convex transformed classes with $h(x) = x$, so that $f(x) = \varphi(x)$ with φ convex (or convex and decreasing).

Example: $f(x) = \exp(-|x|)1_{(0,\infty)^d}(x)$.

A. Some multivariate shape-constrained classes

Log-concave densities on \mathbb{R}^d



$$f(\underline{x}) = \exp(\varphi(\underline{x})) = \exp(-(-\varphi(\underline{x})))$$

where $\varphi : \mathbb{R}^d \mapsto \mathbb{R}$ is concave (so $-\varphi$ is convex).

- Exponentially decaying tails; does not include multivariate t -densities.

A. Some multivariate shape-constrained classes

- s -convex densities and h -convex densities
(Koenker and Mizera; Seregin, Seregin and Wellner)

$$f(\underline{x}) = h(\varphi(\underline{x}))$$

where $\varphi : \mathbb{R}^d \mapsto \mathbb{R}$ is convex, $h : \mathbb{R} \mapsto \mathbb{R}^+$ is decreasing and continuous; e.g. $h_s(u) \equiv (1 + u/s)^{-s}$ with $s > d$.

Larger classes than log-concave: includes multivariate t_n for $d < s \leq n + d$.

B: Review of available theory: MLEs for multivariate classes

Block decreasing densities on \mathbb{R}^{+d}

Problem	Lower Bound	Upper Bound / MLE
A (consist)		Polonik (1995,1998) Pavlidis (2008?)
B (local)	Pavlidis (2008 & 2009) rate: $n^{1/(d+2)}$ $\left\{ \prod_{j=1}^d \frac{\partial f}{\partial x_j}(x) f(x) \right\}^{1/(d+2)}$? rate: $n^{1/(d+2)}$?? const: ??
C (global)	Biau and Devroye (2003) rate: $n^{1/(d+2)}$	Biau and Devroye (2003) analogues of Birgé's histogram estimators

B: Review of available theory: MLEs for multivariate classes

Scale mixtures of uniform on \mathbb{R}^{+d}

Problem	Lower Bound	Upper Bound / MLE
A (consist)		Pavlidis (2008) Pavlidis & W (2010)
B (local)	Pavlidis (2008) rate: $n^{1/3}$ (all d) $\left\{ \frac{\partial^d f}{\partial x_1 \cdots \partial x_d}(x) f(x) \right\}^{1/3}$? Pavlidis (2008), partial results
C (global)	?? (hints from entropy bounds of Blei, Gao, and Li) ??	?? ?? ??

B: Review of available theory: MLEs for multivariate classes

Log concave densities on \mathbb{R}^d

Problem	Lower Bound	Upper Bound / MLE
A (consist)	consistency with misspecification! computation:	Cule and Samworth (2010a) Schumacher, Rufibach, Samworth (2010) Cule, Samworth, Stewart (2010)
B (local)	Seregin (2010) rate: $n^{2/(d+4)}$ $\{f^{d+2}(x)\text{curv}_x(\varphi)\}^{1/(d+4)}$ $\text{curv}_x(\varphi) = \det \nabla^2 \varphi(x)$?? ?? ??
C (global)	?? conjectures: Seregin and W (2010)	?? ?? ??

B: Review of available theory: MLEs for multivariate classes

s -convex and h -convex densities on \mathbb{R}^d

Problem	Lower Bound	Upper Bound / MLE
A (consist)	computation: (related estimators) (convex regression)	Seregin & W (2010) Koenker & Mizera (2010) Seijo and Sen (2010)
B (local)	Seregin (2010) rate: $n^{2/(d+4)}$ $\left\{ \frac{f(x)\text{curv}_x(\varphi)}{h'(\varphi(x))^4} \right\}^{1/(d+4)}$	MLE & LSE rate inefficient $d > 4$?? ??
C (global)	?? ?? ??	?? LSE rate inefficient, $d > 4??$ or $d \geq 4??$

C: Some alternative classes on \mathbb{R} :

Bondesson's hyperbolically monotone classes

Let $k \geq 1$ be an integer. $f : (0, \infty) \rightarrow \mathbb{R}^+$ is *hyperbolically monotone* of order k (HM_k) if, for each fixed $u > 0$, the function

$$H(w) \equiv f(uv)f(u/v), \quad w \equiv \frac{1}{2} \left(v + \frac{1}{v} \right) \geq 1,$$

is such that

$$\begin{aligned} (-1)^j H^{(j)}(w) &\geq 0, \quad \text{for } j = 0, \dots, k-1, \\ (-1)^k H^{(k-1)}(w) &\text{ is right continuous and decreasing.} \end{aligned}$$

If f is hyperbolically monotone for all k , f is said to be *hyperbolically completely monotone* (HCM or HM_∞).

Note that $f(uv)f(u/v)$ is always a function of w by symmetry, and for $v \geq 1$ $v = w + \sqrt{w^2 - 1}$.

C: Some alternative classes on \mathbb{R} :

Bondesson's hyperbolically monotone classes

Example 1. Half-normal distributions

$$f(x) = \sqrt{\frac{2}{\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) \mathbf{1}_{(0,\infty)}(x)$$

has

$$\begin{aligned} H(w) &= f(uv)f(u/v) = \frac{2}{\pi\sigma^2} \exp\left(-\frac{u^2}{2\sigma^2} \left(v + \frac{1}{v}\right)^2 + \frac{u^2}{\sigma^2}\right) \\ &= \frac{2}{\pi\sigma^2} \exp\left(-\frac{u^2}{2\sigma^2} w^2 + \frac{u^2}{\sigma^2}\right). \end{aligned}$$

For fixed $u > 0$ $w \mapsto H(w)$ is decreasing, but $-H'$ is not. Thus $f \in HM_1$ while $f \notin HM_2$.

Example 2. Uniform (a, b) If $f(x) = (b - a)^{-1} \mathbf{1}_{(a,b)}(x)$ with $0 \leq a < b$, then for $v \geq 1$

$$H(w) = f(uv)f(u/v) = (b - a)^{-2} \mathbf{1}_{\{a < u/v \leq uv < b\}}$$

so $H(w) = \mathbf{1}_{\{1 \leq v < V_u \equiv \min\{u/a, b/u\}\}}$. Thus H is decreasing and $f \in HM_1$.

C: Some alternative classes on \mathbb{R} :

Bondesson's hyperbolically monotone classes

Exercise 1. Show that $f(x) = Cx^{\beta-1}(1+cx)^{-\gamma}1_{(0,\infty)}(x)$, with $\beta, \gamma, c \geq 0$ and C a normalizing constant, satisfies $f \in HM_\infty$.

Exercise 2. (log-normal) $f(x) = C\exp(-(\log x - \mu)^2/2\sigma^2)$ satisfies $f \in HM_\infty$.

Exercise 3. $f(x) = (a-x)_+^{\gamma-1}1_{(0,\infty)}(x)$ satisfies $f \in HM_{[\gamma]}$.

Theorem 1. (Bondesson, 1997) If X and Y are independent random variables such that $X \sim f \in HM_k$ and $Y \sim g \in HM_k$, then $XY \sim HM_k$ and $X/Y \sim HM_k$.

Theorem 2. (Bondesson, 1992) $X \sim f \in HM_1$ if and only if $\log X \sim e^x f(e^x)$ is log-concave.

C: Some alternative classes on \mathbb{R} :

Bondesson's hyperbolically monotone classes

Putting these two results together:

- Transform the hyperbolically monotone classes from \mathbb{R}_+ to \mathbb{R} :

$$\begin{aligned}\mathcal{HM}_k \circ \exp &\equiv \{g(x) = e^x f(e^x) : f \in HM_k\} \\ &\equiv \text{log-hyperbolically } k\text{-monotone}\end{aligned}$$

- $\mathcal{HM}_k \circ \exp$ is closed under convolution.
- $\mathcal{HM}_k \circ \exp$ have the same degree of smoothness as the $k+1$ -monotone densities
- $\mathcal{HM}_\infty \circ \exp$ contains the Gaussian distributions (by Exercise 2).

Conclusion: The classes $\mathcal{HM}_k \circ \exp \searrow \mathcal{HM}_\infty \circ \exp$ provide a nice analogue of the k -monotone classes on \mathbb{R}^+ for \mathbb{R} with nice closure properties.

D: Open problems and questions: \mathbb{R}^1

- Are there “natural” switching relations for the k –monotone MLE’s and / or LSE’s?
- More connections to convexity theory?
- Pointwise rates of convergence for the k –monotone MLE’s?
- Rates of convergence under degenerate mixing, $G = \delta_1$?
- Rates of convergence for the MLE’s of G (inverse problems)?
- Global rates of convergence in L_1 and Hellinger metrics, log-concave classes?
- Theory for natural discrete shape-constrained classes?
(Monotone, convex-decreasing, completely monotone,)
- MLE’s for Bondesson’s HM_k classes?

E: Open problems and questions: \mathbb{R}^d

- local rates and global rates for shape constrained estimators in \mathbb{R}^d ?
- Local (pointwise) limiting distribution theory for MLE's and other natural divergence-based estimators?
- When are the MLE's rate (in-)efficient?
Conjecture 1: Block decreasing: inefficient for $d > 2$.
Conjecture 2: Log-concave and s -concave: inefficient for $d > 4$.
- How to penalize or sieve or ... to obtain rate efficient estimators in these classes for higher dimensions?
- Do there exist natural shape-constraints with smoothness > 2 for which MLE's are rate-efficient and which have natural preservation properties under convolution, marginalization, and so forth?

E: Open problems and questions: \mathbb{R}^d

- Faster and more efficient algorithms?
- Faster and more efficient algorithms?!

Je vous remerci!