

**Strong log-concavity of Chernoff's
density:
connections and open problems**



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Seminaire 1, Toulouse: Tuesday

*[http://www.stat.washington.edu/jaw/
RESEARCH/TALKS/talks.html](http://www.stat.washington.edu/jaw/RESEARCH/TALKS/talks.html)*

Seminaire 2, Toulouse: today

Based on joint work with:

- Fadoua Balabdaoui
- Werner Ehm

Seminaire 3, Monday, 5 March:

Nonparametric estimation of log-concave densities

Seminaire 4, Thursday, 8 March:

A local maximal inequality under uniform entropy

Outline

- A Strong - log-concave densities on \mathbb{R}
- B Bondesson's Harmonically Completely Monotone classes: HM_∞
 - ▶ B1. A product representation for the standard normal density
 - ▶ B2. Strong log-concavity and HM_∞
- C: Pólya frequency functions, the class PF_∞ .
 - ▶ C1. Is Chernoff's density strongly log-concave?
 - ▶ C2. Sums of independent exponential random variables
 - ▶ C3. Interpolation: Gaussian to Chernoff and beyond ...
- D: Questions and problems

A. Strong - log-concave densities on \mathbb{R} and \mathbb{R}^d

- $h : \mathbb{R}^d \rightarrow \bar{\mathbb{R}}$ is *strongly convex* if there exists a constant $c > 0$ such that

$$h(\theta x + (1 - \theta)y) \leq \theta h(x) + (1 - \theta)h(y) - \frac{1}{2}c\theta(1 - \theta)\|x - y\|^2$$

for all $x, y \in \mathbb{R}^d$, $\theta \in (0, 1)$.

- Equivalent to convexity of

$$h(x) - \frac{1}{2}c\|x\|^2$$

for some $c > 0$.

- Replacing h by $-\log f$ leads to a definition of *strong log-concavity* of a (density) function: $f : \mathbb{R}^d \rightarrow \bar{\mathbb{R}}$ is strongly log-concave if and only if

$$-\log f(x) - \frac{1}{2}c\|x\|^2$$

is convex for some $c > 0$.

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- Defining $-\log g(x) \equiv -\log f(x) - (1/2)c\|x\|^2$, it is easily seen that f is **strongly log-concave** if and only if

$$f(x) = g(x)\exp(-(1/2)c\|x\|^2)$$

for some $c > 0$ and log-concave function g .

- Thus if $f \in C^2(\mathbb{R}^d)$, a sufficient condition for strong log-concavity is: $\text{Hess}(-\log f)(x) \geq cI_d$ for all $x \in \mathbb{R}^d$ and some $c > 0$ where I_d is the $d \times d$ identity matrix.

B. Bondesson's Harmonically Completely

Monotone classes: HM_∞

Let $\phi(x) = (2\pi)^{-1/2}\exp(-x^2/2)$ be the standard normal density.

Question: Can we write

$$\phi(x) = (1/2)g(x)g(-x) \tag{1}$$

for some “interesting” functions g . Of course this holds almost trivially for g Gaussian (and hence $g \in PF_\infty$) If

$$g(x) = (8\pi)^{1/4}\phi\left(\frac{x}{\sqrt{2}}\right),$$

then (1) holds.

Can (1) hold for (non-Gaussian) $g \in HM_\infty$?

Definition. A density f on $(0, \infty)$ is **hyperbolically completely monotone**, and we write $f \in HM_\infty$ if

$$H(w) \equiv f(uv)f(u/v)$$

is a completely monotone function of $w \equiv (v + v^{-1})/2$ for every u .

Theorem. (Bondesson, 1992). The class HM_∞ consists of all densities of the form

$$f(x) = Cx^{\beta-1}h_1(x)h_2(1/x) \quad (2)$$

where $\beta \in \mathbb{R}$ and

$$h_j(x) = \exp \left\{ -b_j x + \int \log \left(\frac{y+1}{y+x} \right) d\Gamma_j(y) \right\} \quad (3)$$

for $b_j \geq 0$ and measures Γ_j on $[1, \infty)$ satisfying $\int f(1+y)^{-1} d\Gamma_j(y) < \infty$.

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- If $Y \sim f_Y$ on $(0, \infty)$,
then $X \equiv \log Y \sim f_X(x) = f_Y(e^x)e^x$ on \mathbb{R} .
 - If $X \sim f_X$ on \mathbb{R} ,
then $Y = e^X \sim f_Y(y) = f_X(\log y)y^{-1}$ on $(0, \infty)$.
 - Thus if $f_Y \in HM_\infty$, $X \equiv \log Y$ has density f_X of the form

$$f_X(x) = Ce^{\beta x} \tilde{h}_1(x) \tilde{h}_2(-x) \equiv C \tilde{g}_1(x) \tilde{g}_2(-x) \quad (4)$$

where, for $j = 1, 2$,

$$\tilde{g}_j(x) \equiv e^{\beta_j x} \tilde{h}_j(x) = e^{\beta_j x} h_j(e^x), \quad (5)$$

$$\tilde{h}_j(x) = h_j(e^x) \quad (6)$$

$$= \exp \left(-b_j e^x + \int_1^\infty \log \left(\frac{v+1}{v+e^x} \right) d\Gamma_j(v) \right), \quad (7)$$

and $\beta_1 - \beta_2 \equiv \beta$.

Proposition 1. (Bondesson, 1992). The log-normal density on $(0, \infty)$ given by

$$f(y) = C y^{\mu/\sigma^2 - 1} \exp\left(-\frac{(\log y)^2}{2\sigma^2}\right)$$

is hyperbolically completely monotone; i.e. $f \in HM_\infty$.

Alternative statement: $\phi \in \log(HM_\infty)$.

B1. A product representation for the standard normal density

Consequence: (Bondesson, 1992)

$$\begin{aligned}\phi(z) &= \frac{1}{\sqrt{2\pi}} \exp \left(\int_0^\infty \left\{ \log \left(\frac{e^s + 1}{e^s + e^z} \right) + \log \left(\frac{e^s + 1}{e^s + e^{-z}} \right) \right\} ds \right) \\ &= \frac{1}{2} g(z) g(-z)\end{aligned}\tag{8}$$

where

$$\begin{aligned}g(z) &\equiv (2/\pi)^{1/4} \exp(z) \exp \left(\int_0^\infty \log \left(\frac{e^s + 1}{e^s + e^z} \right) ds \right) \\ &= (2/\pi)^{1/4} \exp \left(\frac{\pi^2}{12} + z + \int_{-e^z}^0 \frac{\log(1-t)}{t} dt \right)\end{aligned}\tag{9}$$

is log-concave, integrable, and $g \in \log(HM_\infty)$.

What is the measure $\Gamma = \Gamma_1 = \Gamma_2$? In fact $d\Gamma(y) = y^{-1} dy!$

Is this g in PF_∞ ?

Proof of Proposition 1. Note that

$$\begin{aligned}
\frac{1}{y} \left\{ \frac{1}{x+y} - \frac{1}{x^2 y + x^{-1}} \right\} &= \frac{1}{y} \left\{ \frac{x^2(y + x^{-1}) - (y + x)}{x^2(y + x)(y + x^{-1})} \right\} \\
&= \frac{x^2 - 1}{x^2(y + x)(y + x^{-1})} = \frac{x^2 - 1}{x^2(1 + xy)(1 + y/x)} \\
&= \frac{1}{(1 + xy)(1 + y/x)} - \frac{1/x^2}{(1 + xy)(1 + y/x)} \\
&= \frac{1}{x} \left\{ \frac{1}{y + x^{-1}} - \frac{1}{y + x} \right\}.
\end{aligned}$$

Thus it follows that

$$\begin{aligned}
\int_1^\infty \left\{ \frac{1}{x+y} - \frac{1}{x^2 y + x^{-1}} \right\} \frac{1}{y} dy &= \frac{1}{x} \int_1^\infty \left\{ \frac{1}{y + x^{-1}} - \frac{1}{y + x} \right\} dy \\
&= \frac{1}{x} \{ \log(y + 1/x) - \log(y + x) \} \Big|_1^\infty = \frac{1}{x} \log x \quad (10)
\end{aligned}$$

Also note that the integral on the right side in (10) can be rewritten as

$$\begin{aligned}
 \int_1^\infty \left\{ \frac{1}{y+x^{-1}} - \frac{1}{y+x} \right\} dy &= \int_1^\infty \frac{y+x - (y+x^{-1})}{(y+x^{-1})(y+x)} dy \\
 &= (x-x^{-1}) \int_1^\infty \frac{1}{y^2 + (x+x^{-1})y + 1} dy = (x-x^{-1}) \frac{x \log x}{x^2 - 1} \\
 &= \log x;
 \end{aligned}$$

this rewrite makes the integrability completely clear. Integrating the resulting identity

$$\frac{\log x}{x} = \int_1^\infty \left\{ \frac{1}{x+y} - \frac{1}{x^2} \frac{1}{y+x^{-1}} \right\} \frac{1}{y} dy$$

with respect to x on both sides over $[1, z]$ and using Fubini's theorem

(permitted because the integrand is non-negative) yields

$$\begin{aligned}
 \frac{1}{2}(\log z)^2 &= \int_1^z \left\{ \int_1^\infty \left\{ \frac{1}{x+y} - \frac{1}{x^2} \frac{1}{y+x^{-1}} \right\} \frac{1}{y} dy \right\} dx \\
 &= \int_1^\infty \left\{ \int_1^z \left\{ \frac{1}{x+y} - \frac{1}{x^2} \frac{1}{y+x^{-1}} \right\} dx \right\} \frac{1}{y} dy \\
 &= \int_1^\infty \left\{ \log(y+x) + \log(y+x^{-1}) \right\} \Big|_1^z \frac{1}{y} dy \\
 &= - \int_1^\infty \left\{ \log \left(\frac{y+1}{y+z} \right) + \log \left(\frac{y+1}{y+z^{-1}} \right) \right\} \frac{1}{y} dy.
 \end{aligned}$$

Making the change of variable of integration $y = e^s$ gives

$$\frac{1}{2}(\log z)^2 = - \int_0^\infty \left\{ \log \left(\frac{e^s + 1}{e^s + z} \right) + \log \left(\frac{e^s + 1}{e^s + z^{-1}} \right) \right\} ds,$$

Changing the variable z to e^x then yields the claim:

$$-\frac{1}{2}x^2 = \int_0^\infty \left\{ \log \left(\frac{e^s + 1}{e^s + e^x} \right) + \log \left(\frac{e^s + 1}{e^s + e^{-x}} \right) \right\} ds.$$

The claimed identity involving g follows by direct substitution. To see the second form of g , note that

$$\begin{aligned}
 \int_1^\infty \log\left(\frac{y+1}{y+z}\right) \frac{1}{y} dy &= \int_1^\infty \log\left(\frac{1+1/y}{1+z/y}\right) \frac{1}{y} dy \\
 &= \int_0^1 \log\left(\frac{1+t}{1+tz}\right) \frac{1}{t} dt \\
 &\quad \text{by the change of variables } t = 1/y \\
 &= \int_0^1 \log(1+t) \frac{1}{t} dt - \int_0^1 \log(1+tz) \frac{1}{t} dt \\
 &= \frac{\pi^2}{12} + \int_{-z}^0 \log(1-s) \frac{1}{s} ds \\
 &\quad \text{by the change of variable } -s = tz.
 \end{aligned}$$

To see that g is log-concave, note that

$$-\log g(z) = -\frac{\pi^2}{12} - z - \int_{-e^z}^0 \frac{\log(1-t)}{t} dt,$$

and hence

$$(-\log g)'(z) = -1 + \frac{\log(1 + e^z)}{-e^z} \cdot (-e^z) = -1 + \log(1 + e^z),$$

$$(-\log g)''(z) = \frac{e^z}{1 + e^z} \geq 0.$$

Integrability of g follows from

$$-\log g(z) \sim \begin{cases} -z & \text{as } z \rightarrow -\infty, \\ -z + z^2/2 & \text{as } z \rightarrow \infty, \end{cases}$$

where we used

$$\int_{-y}^0 \frac{\log(1-t)}{t} dt \sim \begin{cases} 0 & \text{as } y \searrow 0, \\ -(1/2)(\log y)^2 & \text{as } y \rightarrow \infty. \end{cases}$$

B2. Strong log-concavity and HM_∞

Sufficient conditions for $f \in \log(HM_\infty)$ to be strongly log-concave?

Proposition: Suppose that f_X is given by (4) - (6).

A. If $b_1 > 0$ and $b_2 > 0$, then f_X is strongly log-concave for any (all) measures Γ_1 and Γ_2 .

B. Suppose $b_1 = 0$ or $b_2 = 0$ and $d\Gamma_j(v) = v^{-1}r_j(v)dv$ for $j = 1, 2$ where (at least one of) r_1 and r_2 satisfy:

- (i) $r_j(y) \geq 0$ all $y \in [1, \infty)$ with strict inequality for some $y > 0$;
- (ii) r_j is non-decreasing.

Then $v_j(x) \equiv (-\log g_j)''(x) \geq v_j(0) > 0$ for all $x \geq 0$ and hence f_X is strongly log-concave with $(-\log f_X)''(x) \geq \max\{v_1(0), v_2(0)\} > 0$.

Proof of Proposition: Since Γ_j has density $\gamma_j(y) = y^{-1}r_j(y)$,

$$\begin{aligned}
(-\log g_j)''(x) &= \int_0^\infty \frac{e^{w-x}}{(1+e^{w-x})^2} e^w \gamma_j(e^w) dw \\
&= \int_0^\infty \frac{e^{w-x}}{(1+e^{w-x})^2} r_j(e^w) dw \quad \text{if } \gamma(y) = y^{-1}r_j(y) \\
&= \int_{-x}^\infty \frac{e^z}{(1+e^z)^2} r_j(e^{z+x}) dz \\
&= Er_j(e^{Z+x}) \mathbf{1}\{Z \geq -x\} \equiv v(x)
\end{aligned}$$

where $Z \sim$ standard logistic with density $e^z/(1+e^z)^2$. Then it follows that

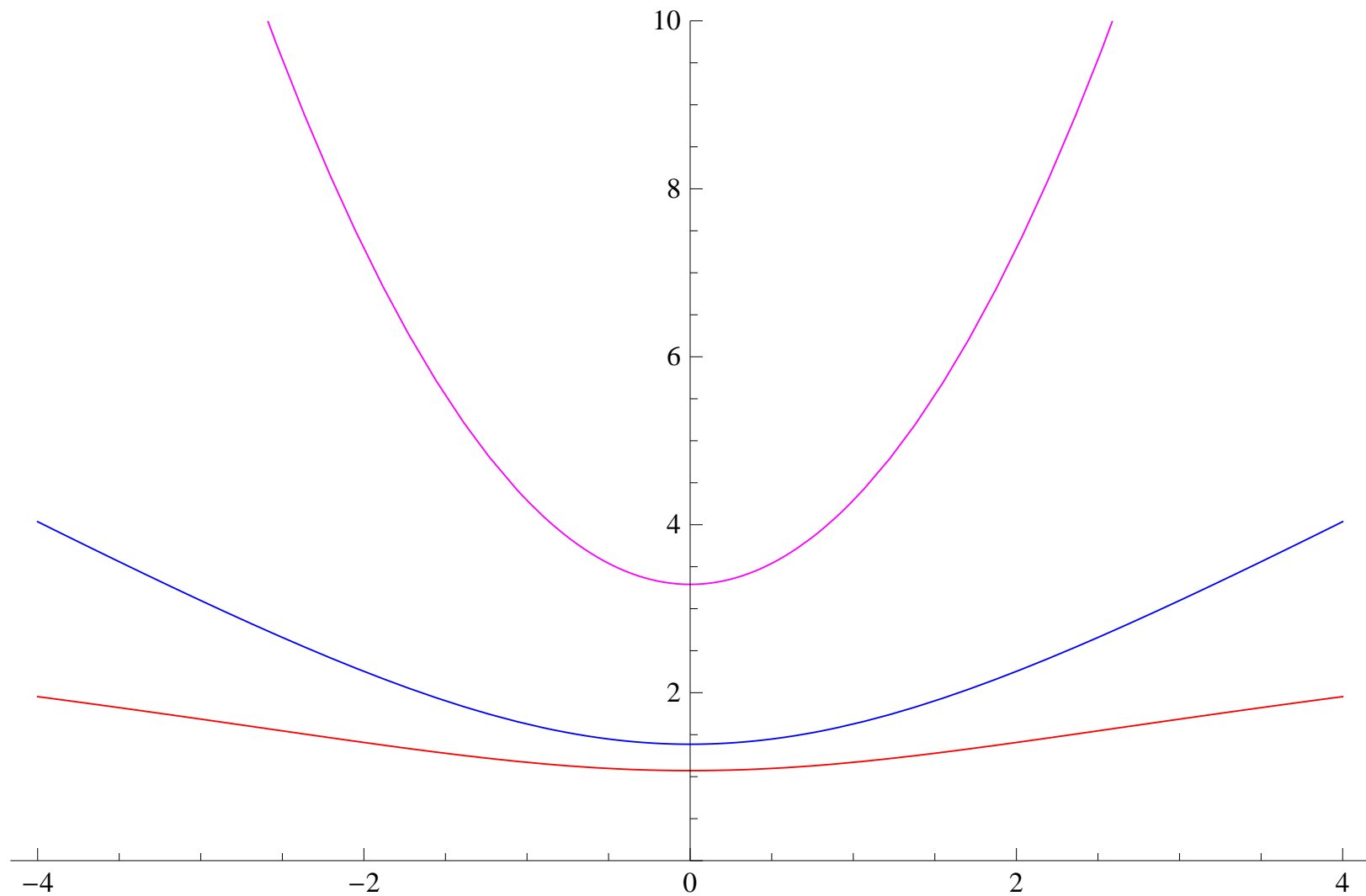
$$\begin{aligned}
v_j(x) &= Er_j(e^{Z+x}) \mathbf{1}\{Z+x \geq 0\} \\
&= Er_j(e^{Z+x}) \mathbf{1}_{[-x,0]}(Z) + Er_j(e^{Z+x}) \mathbf{1}_{(0,\infty)}(Z) \\
&\geq Er_j(e^{Z+x}) \mathbf{1}_{(0,\infty)}(Z) \geq Er_j(e^Z) \mathbf{1}_{(0,\infty)}(Z) = v_j(0) > 0.
\end{aligned}$$

Thus the density f_X is strongly log-concave.

Example. $b_1 = b_2 = 0$ and $r(y) = (\log y)^c$ with $c > 0$; thus $d\Gamma(y) = y^{-1}(\log y)^c dy$. Then

$$\begin{aligned}(-\log g)''(x) &\equiv v(x) \\ &= \int_1^\infty \frac{ve^x}{(v + e^x)^2} d\Gamma(v) \equiv v(x) \\ &= Er(e^{Z+x})1\{Z \geq -x\} \\ &= E(Z+x)^c 1\{Z \geq -x\} \sim x^c \text{ as } x \rightarrow \infty.\end{aligned}$$

$$w(x) \equiv v(x) + v(-x) = (-\log f)''(x) \quad \text{for } f(x) = (1/2)g(x)g(-x).$$



$w(x) = (-\log f)''(x)$, $c = 1/2$ (red), $c = 1$ (blue), $c = 2$ (purple)

C: Pólya frequency functions, the class PF_∞ .

C1: Is Chernoff's density strongly log-concave?

Recall that From Chernoff (1964)

$$f(z) = \frac{1}{2}g(z)g(-z) \equiv \frac{1}{2}g_1(z)g_1(-z).$$

From Groeneboom (1885, 1989), Daniels and Skyrme (1985)

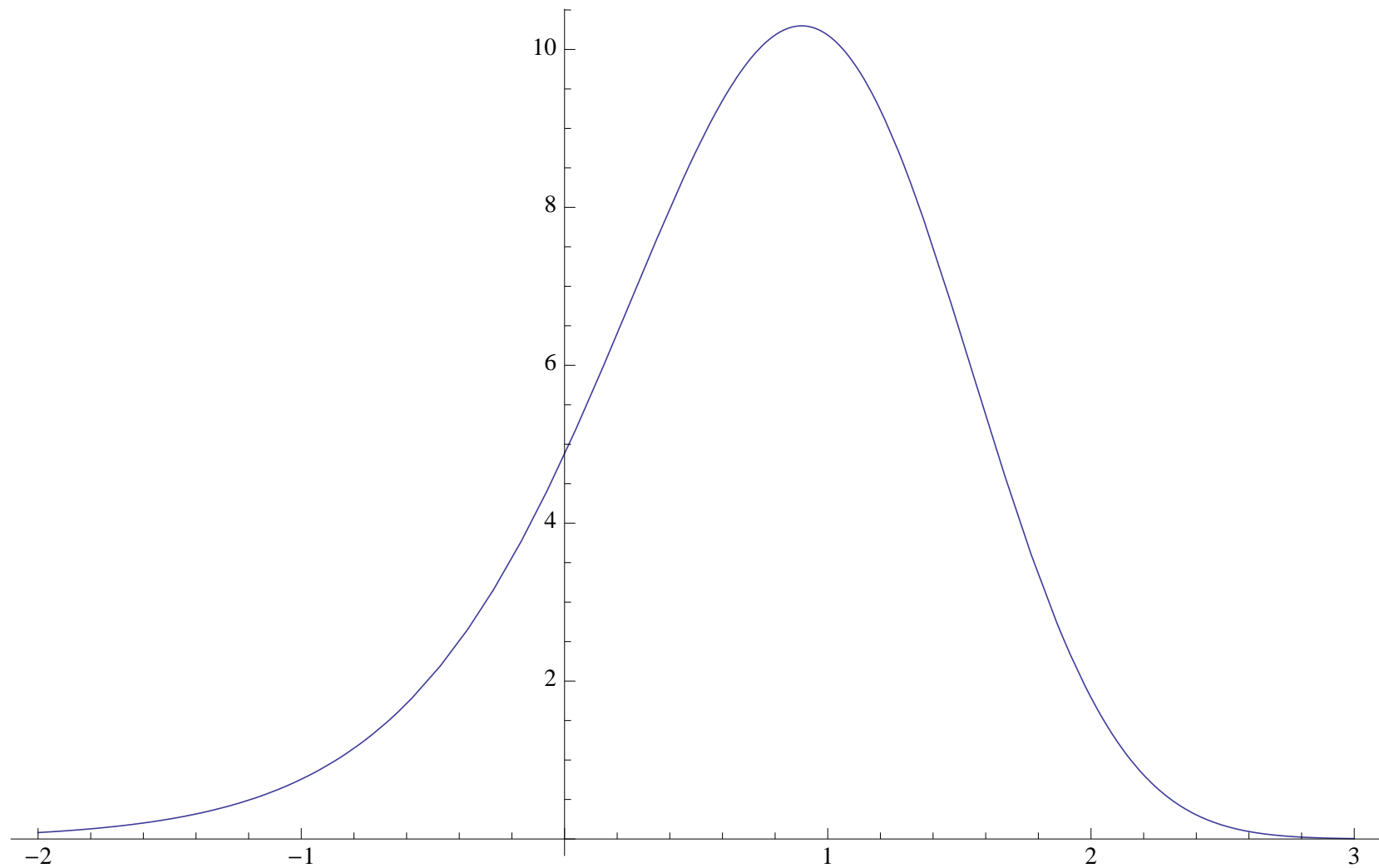
$$\hat{g}_c(\lambda) = \int_{-\infty}^{\infty} e^{i\lambda s} g_c(s) ds = \frac{2^{1/3} c^{-1/3}}{Ai(i(2c^2)^{-1/3}\lambda)}.$$

From Salmassi (1999) and Merkes and Salmassi (1997)

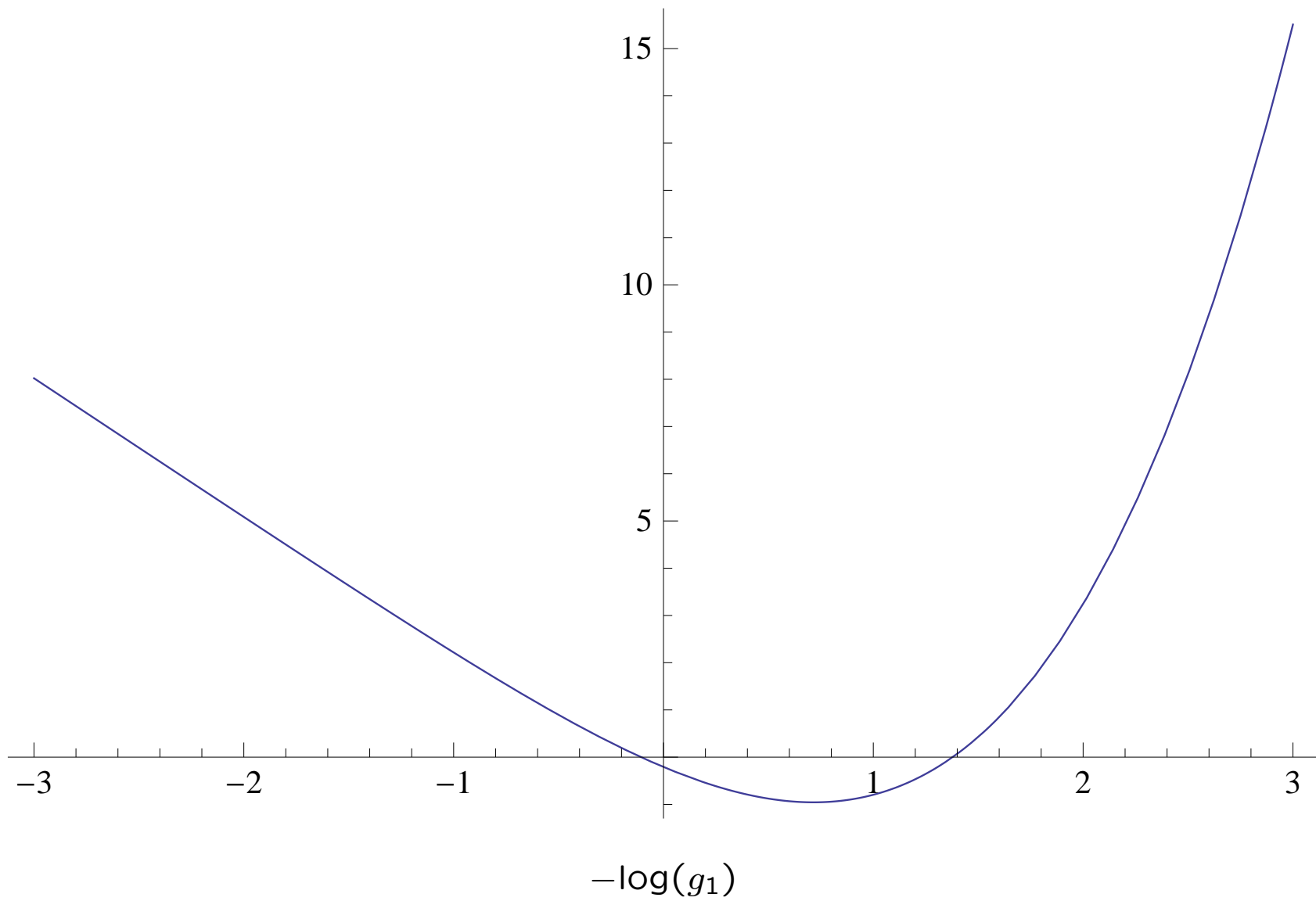
$$Ai(z) = Ai(0)e^{-\nu z} \prod_{j=1}^{\infty} (1 + z/a_j)\exp(-z/a_j)$$

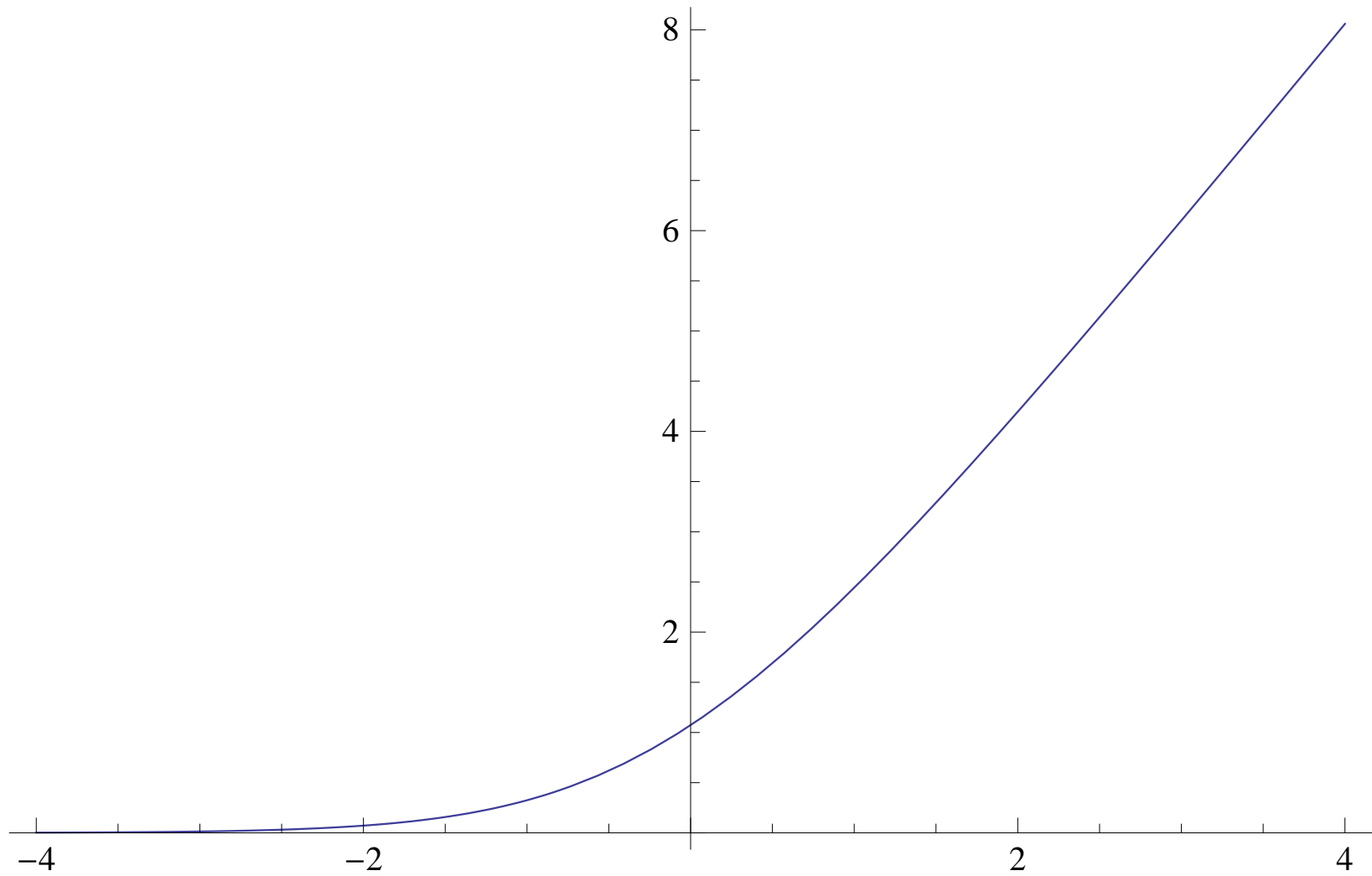
where $\{-a_k\}$ are the zeros of the Airy function Ai (so that $a_k > 0$ for each k). where $Ai(0) = c_1 = 1/(3^{2/3}\Gamma(2/3)) \approx 0.35503$, and

$$\nu = -Ai'(0)/Ai(0) = 3^{1/3}\Gamma(2/3)/\Gamma(1/3) \approx .729011 \dots$$



The function g_1





C2. Sums of independent exponential random variables

Suppose that $X \sim \text{Exp}(\lambda)$. Then $f(x) = \lambda e^{-\lambda x} \mathbf{1}_{(0, \infty)}(x)$ and

$$\begin{aligned} Ee^{sX} &= \int_0^{\infty} e^{sx} \lambda e^{-\lambda x} dx = \lambda \int_0^{\infty} e^{-(\lambda-s)x} dx \\ &= \frac{\lambda}{\lambda-s} = \frac{1}{1-\frac{s}{\lambda}} \end{aligned}$$

for $s < \lambda$. Thus

$$E \exp(s(X - 1/\lambda)) = \exp(-s/\lambda) \frac{1}{1-s/\lambda} = \frac{1}{(1-s/\lambda)e^{s/\lambda}}.$$

Now suppose that $X_i \sim \text{Exp}(1/b_i)$ for $i = 1, 2, \dots$ are independent. Thus

$$E \exp(-s(X_i - b_i)) = \frac{1}{(1 + b_i s)e^{-b_i s}},$$

and

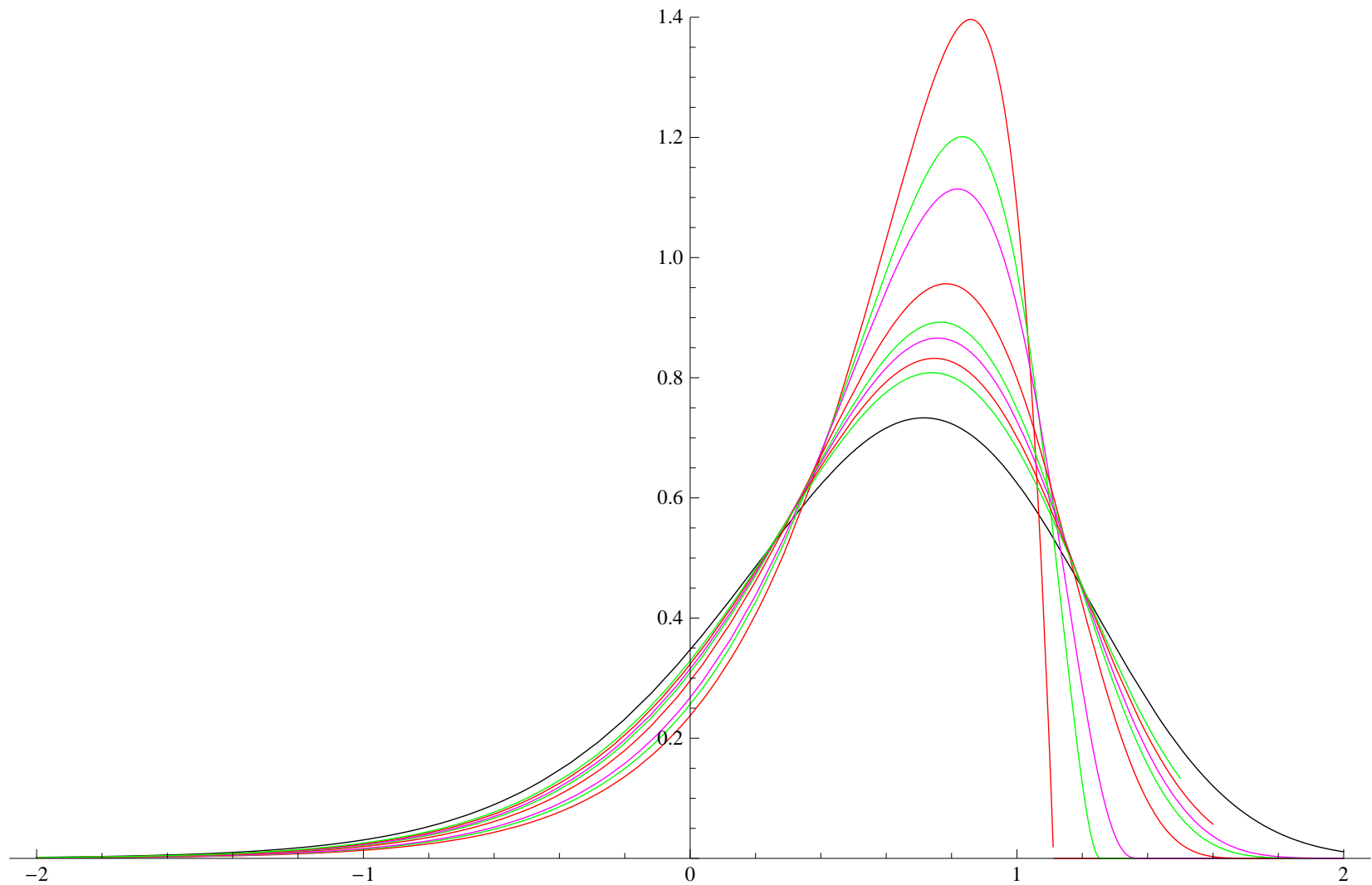
$$\begin{aligned} E \exp\left(-s \sum_{i=1}^{\infty} (X_i - b_i)\right) &= E \left\{ \prod_{i=1}^{\infty} e^{-s(X_i - b_i)} \right\} = \prod_{i=1}^{\infty} E e^{-s(X_i - b_i)} \\ &= \frac{1}{\prod_{i=1}^{\infty} (1 + b_i s) e^{-b_i s}}. \end{aligned}$$

This is exactly the form of one term of the reciprocal of the bilateral Laplace or Fourier transform in Schoenberg's theorem characterizing PF_{∞} if we require $\sum_1^{\infty} b_j^2 < \infty$. In the context of Chernoff's distribution $b_j = 1/a_j$ where $\{-a_j\}$ are the zeros of the Airy function, with $a_j \sim ((3/2)\pi(k - 1/4))^{2/3}$.

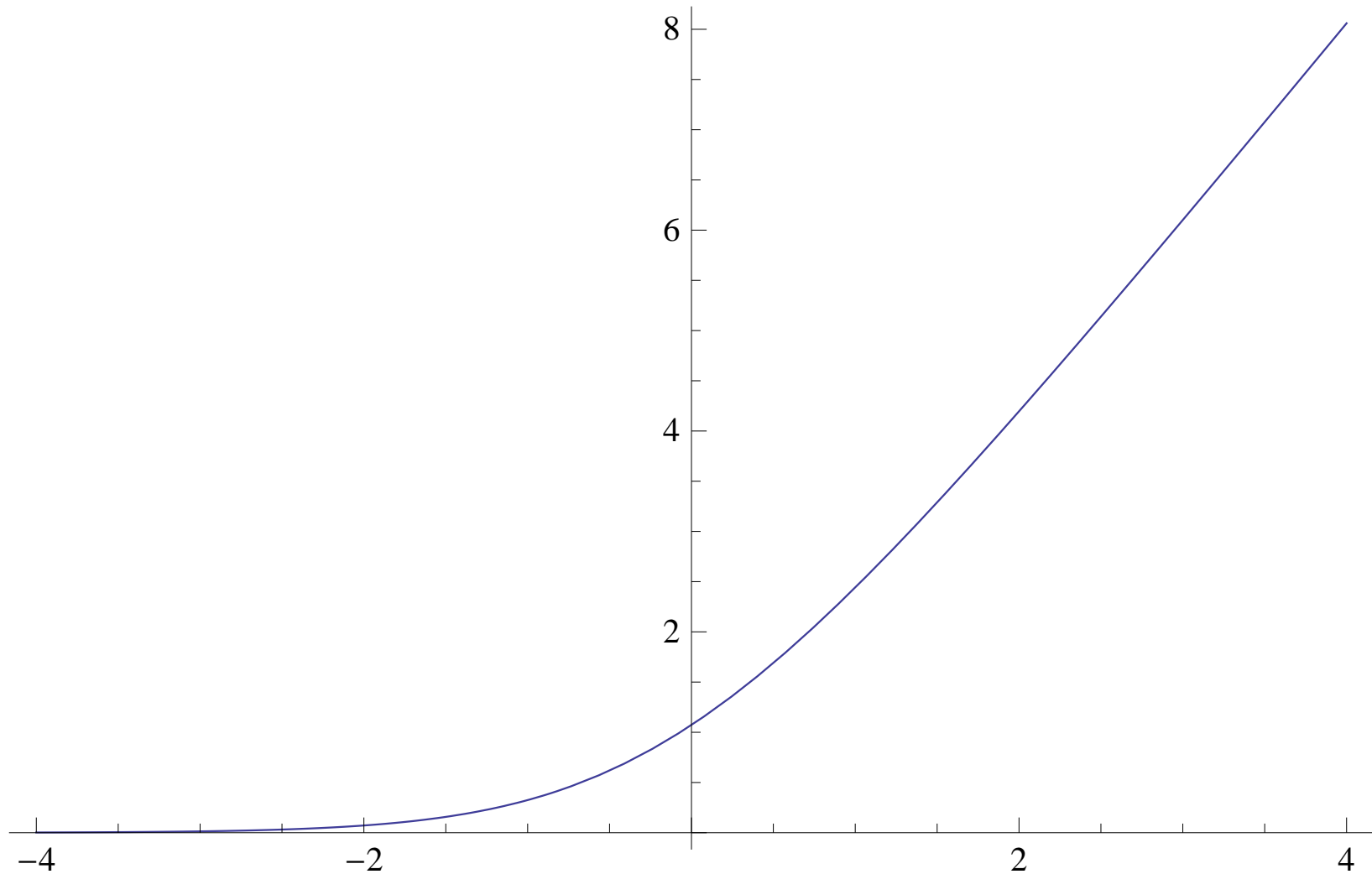
This leads to consideration of the distributions of

$$\sum_{j=1}^m (X_j - b_j) \quad \text{and} \quad \sum_{j=1}^m X_j \equiv Y_m.$$

What do we know about the distribution of Y_m ?



Finite m approximations of \tilde{g} , $m \in \{2, 3, 4, 10, 20, 30, 60, 120\}$



Second derivative, $(-\log(g_1))''(x)$

Proposition 1. (Harrison, 1990) If $X_i \sim \text{Exp}(1/b_i) \equiv \text{Exp}(\lambda_i)$ for $i \geq 1$ be independent. Then the density f_m of Y_m is given by

$$f_m(t) = \sum_{j=1}^m \lambda_j \exp(-\lambda_j t) \prod_{i \neq j} \frac{\lambda_i}{\lambda_i - \lambda_j} 1_{(0, \infty)}(t).$$

Define

$$v_m(t) \equiv (-\log f_m)''(t).$$

Conjecture: v_m is convex (and decreasing) for every m .

Proposition 2. v_2 is convex.

Proof: By Harrison's formula

$$\begin{aligned} f_2(x) &= \lambda_1 e^{-\lambda_1 x} \frac{\lambda_2}{\lambda_2 - \lambda_1} + \lambda_2 e^{-\lambda_2 x} \frac{\lambda_1}{\lambda_1 - \lambda_2} \\ &= \frac{\lambda_1 \lambda_2}{\lambda_2 - \lambda_1} e^{-\lambda_1 x} \left(1 - e^{-(\lambda_2 - \lambda_1)x} \right), \end{aligned}$$

and hence

$$-\log f_2(x) = -\log \left(\frac{\lambda_1 \lambda_2}{\lambda_2 - \lambda_1} \right) + \lambda_1 x - \log \left(1 - e^{-(\lambda_2 - \lambda_1)x} \right).$$

Thus we compute

$$\begin{aligned} (-\log f_2)'(x) &= \lambda_1 - \frac{(\lambda_2 - \lambda_1)e^{-(\lambda_2 - \lambda_1)x}}{1 - e^{-(\lambda_2 - \lambda_1)x}} \\ &= \lambda_1 + (\lambda_2 - \lambda_1) \left(1 - \frac{1}{1 - e^{-(\lambda_2 - \lambda_1)x}} \right), \end{aligned}$$

and

$$(-\log f_2)''(x) = (\lambda_2 - \lambda_1)^2 \frac{e^{-(\lambda_2 - \lambda_1)x}}{(1 - e^{-(\lambda_2 - \lambda_1)x})^2} \geq 0$$

for all $x > 0$ (the ≥ 0 part we know already from log-concavity of the two (different) exponential densities and preservation of log-concavity under convolution). Now set $c \equiv \lambda_2 - \lambda_1$ and consider

$$h(x) \equiv h_c(x) = e^{-cx} / (1 - e^{-cx})^2.$$

Now

$$h'(x) = -ce^{-cx} \left\{ \frac{1 - e^{-cx} + 2e^{-cx}}{(1 - e^{-cx})^3} \right\} = -ce^{-cx} \frac{1 + e^{-cx}}{(1 - e^{-cx})^3}.$$

Thus h is decreasing. Furthermore,

$$\begin{aligned} h''(x) &= c^2 e^{-cx} \frac{1 + e^{-cx}}{(1 - e^{-cx})^3} + c^2 e^{-2cx} \frac{1}{(1 - e^{-cx})^3} + 3c^2 e^{-2cx} \frac{1 + e^{-cx}}{(1 - e^{-cx})^4} \\ &= c^2 e^{-cx} \frac{1 + 4e^{-cx} + e^{-2cx}}{(1 - e^{-cx})^4} \\ &> 0. \end{aligned}$$

Thus $(-\log f_2)''$ is convex. □

C3. Interpolation: Gaussian to Chernoff and beyond ...

Question: Does there exist a natural family of densities f_γ with $\gamma \in [1/2, 1)$ satisfying:

- $f_{1/2} = \phi$ and $f_{2/3} =$ Chernoff's density f_1 .
- f_γ is strongly log-concave for each γ .
- With $\beta \equiv 1 - \gamma$ we have

$$(-\log f_\gamma)''(x) \sim \left(\frac{1}{\beta} - 1\right) x^{\frac{1}{\beta}-2} \quad \text{as } x \rightarrow \infty.$$

-

$$f_\gamma(x) \sim D_\gamma x^\kappa \exp(-\beta x^{1/\beta}) \quad \text{as } x \rightarrow \infty.$$

where $\kappa = \kappa_\gamma = (1 - 2\beta)/\beta$.

Such a family would **interpolate** between the Gaussian density ϕ (for $\gamma = 1/2$), Chernoff's density (for $\gamma = 2/3$), and yield increasingly light-tailed densities in the range $2/3 < \gamma < 1$. It

C3. Interpolation: Gaussian to Chernoff and beyond ...

would also correspond to a “full range” of the densities of the form

$$f(x) = \frac{1}{2}g(x)g(-x) \quad \text{with} \quad g \in PF_\infty$$

On-going work with Werner Ehm ... !

D: Questions and problems

- Is Chernoff's density strongly log-concave?
- is $v_m \equiv (-\log f_m)''$ convex for every m and $0 < \lambda_1 < \lambda_2 < \dots$?
- Is there a “natural family” interpolating between standard normal and Chernoff?

- Are the densities corresponding to limit distributions for nonparametric estimators of convex functions (as will be discussed in my talk on Monday) log-concave or strongly log-concave?

Merci!



