# A note on Talagrand's convex hull concentration inequality

## David Pollard[1]

*Yale University*

**Abstract:** The paper reexamines an argument by Talagrand that leads to a remarkable exponential tail bound for the concentration of probability near a set. The main novelty is the replacement of a mysterious calculus inequality by an application of Jensen's inequality.

## 1. Introduction

Let $\mathcal{X}$ be a set equipped with a sigma-field $\mathcal{A}$. For each vector $w = (w_1, \ldots, w_n)$ in $\mathbb{R}^n_+$, the weighted Hamming distance between two vectors $x = (x_1, \ldots, x_n)$ and $y = (y_1, \ldots, y_n)$, in $\mathcal{X}^n$ is defined as

$$d_w(x, y) := \sum_{i \leq n} w_i h_i(x, y) \qquad \text{where } h_i(x, y) = \begin{cases} 1 & \text{if } x_i \neq y_i \\ 0 & \text{otherwise.} \end{cases}$$

For a subset $A$ of $\mathcal{X}^n$ and $x \in \mathcal{X}^n$, the distances $d_w(x, A)$ and $D(x, A)$ are defined by

$$d_w(x) := \inf\{y \in A : d_w(x, y)\} \quad \text{and} \quad D(x, A) := \sup_{w \in \mathcal{W}} d_w(X, A),$$

where the supremum is taken over all weights in the set

$$\mathcal{W} := \left\{ (w_1, \ldots, w_n) : w_i \geq 0 \text{ for each } i \text{ and } |w|^2 := \sum_{i \leq n} w_i^2 \leq 1 \right\}.$$

Talagrand ([10], Section 4.1) proved a remarkable concentration inequality for random elements $X = (X_1, \ldots, X_n)$ of $\mathcal{X}^n$ with independent coordinates and subsets $A \in \mathcal{A}^n$:

$$(1) \qquad \mathbb{P}\{X \in A\}\mathbb{P}\{D(X, A) \geq t\} \leq \exp(-t^2/4) \qquad \text{for all } t \geq 0.$$

As Talagrand showed, this inequality has many applications to problems in combinatorial optimization and other areas. See [12], [7], Chapter 6 and [6], Section 4, for further examples.

There has been a strong push in the literature to establish concentration and deviation inequalities by "more intuitive" methods, such as those based on the tensorization, as in [1, 3–5]. I suspect the search for alternative approaches has been driven by the miraculous roles played by some elementary inequalities in Talagrand's proofs.

[1]Statistics Department, Yale University, Box 208290 Yale Station, New Haven, CT 06520-8290, USA, e-mail: david.pollard@yale.edu; url: http://www.stat.yale.edu/~pollard/

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Talagrand [10] used an induction on $n$ to establish his result. He invoked a slightly mysterious inequality in the inductive step,

$$\inf_{0 \leq \theta \leq 1} u^{-\theta} \exp\left(\frac{(1-\theta)^2}{4}\right) \leq 2 - u \qquad \text{for } 0 < u < 1,$$

which he borrowed from [2] – see Talagrand's discussion following his Lemma 4.1.3 for an explanation of how those authors generalized the earlier result by [8]. There is similar mystery in Talagrand's Lemma 4.2.1, which (in my notation) asserts that

$$\sup_{0 \leq \theta \leq 1} u^{-\theta/c} \exp\left(\psi_c(1-\theta)\right) \leq \frac{1 + c - u}{c} \qquad \text{for } 0 < u < 1,$$

where $\psi_c$ is defined in equation (6) below. (My $\psi_c(u)$ equals Talagrand's $\xi(\alpha, u)$ with $\alpha = 1/c$.) [9] had used this inequality to further generalize the result of [2], giving the concentration inequality listed as Theorem 4.2.4 [10]. It was my attempts to understand how he arrived at his $\xi(\alpha, u)$ function that led me to the conccavity argument that I present in the present note.

It is my purpose to modify Talagrand's proof so that the inductive step becomes a simple application of the Hölder inequality (essentially as in the original proof) and the Jensen inequality. Most of my methods are minor variations on the methods in the papers just cited; my only claim of originality is for the recognition that the mysterious inequalities can be replaced by more familiar appeals to concavity. See also the Remarks at the end of this Section.

The distance $D(x, A)$ has another representation, as a minimization over a convex subset of $[0, 1]$. Write $h(x, y)$ for the point of $\{0, 1\}^n$ with $i$th coordinate $h_i(x, y)$. For each fixed $x$, the function $h(x, \cdot)$ maps $A$ onto a subset $h(x, A) := \{h(x, y) : y \in A\}$ of $\{0, 1\}^n$. The convex hull, $\text{co}\,(h(x, A))$, of $h(x, A)$ in $[0, 1]^n$ is compact, and

$$D(x, A) = \inf\{|\xi| : \xi \in \text{co}\,(h(x, A))\}.$$

Each point $\xi$ of $\text{co}\,(h(x, A))$ can be written as $\int h(x, y)\,\nu(dy)$ for a $\nu$ in the set $\mathcal{P}(A)$ of all Borel probability measures for which $\nu(A) = 1$. That is, $\xi_i = \nu\{y \in A : y_i \neq x_i\}$. Thus

$$(2) \qquad D(x, A)^2 = \inf_{\nu \in \mathcal{P}(A)} \sum_{i \leq n} \left(\nu\{y \in A : y_i \neq x_i\}\right)^2.$$

Talagrand actually proved inequality (1) by showing that

$$(3) \qquad \mathbb{P}\{X \in A\}\mathbb{P}\exp\left(\tfrac{1}{4}D(X, A)^2\right) \leq 1.$$

He also established an even stronger result, in which the $D(X, A)^2/4$ in (3) is replaced by a more complicated distance function.

For each convex, increasing function $\psi$ with $\psi(0) = 0 = \psi'(0)$ define

$$(4) \qquad F_\psi(x, A) := \inf_{\nu \in \mathcal{P}(A)} \sum_{i \leq n} \psi\left(\nu\{y \in A : y_i \neq x_i\}\right),$$

For each $c > 0$ ([10], Section 4.2) showed that

$$(5) \qquad (\mathbb{P}\{X \in A\})^c \mathbb{P}\exp\left(F_{\psi_c}(X, A)\right) \leq 1,$$

where

$$\psi_c(\theta) := c^{-1}\left((1-\theta)\log(1-\theta) - (1-\theta+c)\log\left(\frac{(1-\theta)+c}{1+c}\right)\right)$$

(6) $$= \sum_{k\geq 2}\frac{\theta^k}{k}\left(\frac{R_c + R_c^2 + \cdots + R_c^{k-1}}{(k-1)}\right) \text{ with } R_c := \frac{1}{c+1}.$$

$$\geq \frac{\theta^2}{2+2c}$$

As you will see in Section 3, this strange function is actually the largest solution to a differential inequality,

$$\psi''(1-\theta) \leq 1/(\theta^2 + \theta c) \qquad \text{for } 0 < \theta < 1.$$

Inequality (5) improves on (3) because $D(x, A)^2/4 \leq F_{\psi_1}(x, A)$.

Following the lead of [10], Section 4.4, we can ask for general conditions on the convex $\psi$ under which an analog of (5) holds with some other decreasing function of $\mathbb{P}\{X \in A\}$ as an upper bound. The following slight modification of Talagrand's theorems gives a sufficient condition in a form that serves to emphasize the role played by Jensen's inequality.

**Theorem 1.** *Suppose $\gamma$ is a decreasing function with $\gamma(0) = \infty$ and $\psi$ is a convex function. Define $G_\psi(\eta, \theta) := \psi(1-\theta) + \theta\eta$ and $G_\psi(\eta) := \inf_{0\leq\theta\leq 1} G_\psi(\eta, \theta)$ for $\eta \in \mathbb{R}^+$. Suppose*

    *(i) $r \mapsto \exp(G_\psi(\gamma(r) - \gamma(r_0)))$ is concave on $[0, r_0]$, for each $r_0 \leq 1$*
    *(ii) $(1-p)e^{\psi(1)} + p \leq e^{\gamma(p)}$ for $0 \leq p \leq 1$.*

*Then $\mathbb{P}\exp(F_\psi(X, A)) \leq \exp(\gamma(\mathbb{P}\{X \in A\}))$ for every $A \in \mathcal{A}^n$ and every random element $X$ of $\mathcal{X}^n$ with independent components.*

The next lemma, a more general version of which is proved in Section 3, leads to a simple sufficient condition for the concavity assumption (i) of Theorem 1 to hold.

**Lemma 2 (Concavity lemma).** *Suppose $\psi : [0, 1] \to \mathbb{R}^+$ is convex and increasing, with $\psi(0) = 0 = \psi'(0)$ and $\psi''(\theta) > 0$ for $0 < \theta < 1$. Suppose $\xi : [0, r_0] \to \mathbb{R}^+ \cup \{\infty\}$ is continuous and twice differentiable on $(0, r_0)$. Suppose also that there exists some finite constant $c$ for which $\xi''(r) \leq c\xi'(r)^2$ for $0 < r < r_0$. If*

$$\psi''(1-\theta) \leq 1/(\theta^2 + \theta c) \qquad \text{for } 0 < \theta < 1$$

*then the function $r \mapsto \exp(G_\psi(\xi(r)))$ is concave on $[0, r_0]$.*

Lemma 2 will be applied with $\xi(r) = \gamma(r) - \gamma(r_0)$ for $0 \leq r \leq r_0$. As shown in Section 3, the conditions of the Lemma hold for $\psi(\theta) = \theta^2/4$ with $\gamma(r) = \log(1/r)$ and also for the $\psi_c$ from (6) with $\gamma(r) = c^{-1}\log(1/r)$.

**Remarks.**

    (i) If $\gamma(0)$ were finite, the inequality asserted by Theorem 1 could not hold for all nonempty $A$ and all $X$. For example, if each $X_i$ had a nonatomic distribution and $A$ were a singleton set we would have $F_\psi(X, A) = n\psi(1)$ almost surely. The quantity $\mathbb{P}\exp(F_\psi(X, A))$ would exceed $\exp(\gamma(0))$ for large enough $n$. It it to avoid this difficulty that we need $\gamma(0) = \infty$.

(ii) Assumption (ii) of the Theorem, which is essentially an assumption that the asserted inequality holds for $n = 1$, is easy to check if $\gamma$ is a convex function with $\gamma(1) \geq 0$. For then the function $B(p) := \exp(\gamma(p))$ is convex with $B(1) \geq 1$ and $B'(1) = \gamma'(1)e^{\gamma(1)}$. We have

$$B(p) \geq (1-p)e^{\psi(1)} + p \qquad \text{for all } p \text{ in } [0,1]$$

if $B'(1) \leq 1 - e^{\psi(1)}$.

(iii) A helpful referee has noted that both my specific examples are already covered by Talagrand's results. He (or she) asked whether there are other $(\psi, \gamma)$ pairs that lead to other useful concentration inequalities. A good question, but I do not yet have any convincing examples. Actually, I had originally thought that my methods would extend to the limiting case where $c$ tends to zero, leading to an answer to the question posed on page 128 by [10]. Unfortunately my proof ran afoul of the requirement $\gamma(0) = \infty$. I suspect more progress might be made by replacing the strong assumption on $\psi''$ from Lemma 2 by something closer to the sufficient conditions presented in Section 3.

## 2. Proof of Theorem 1

Argue by induction on $n$. As a way of keeping the notation straight, replace the subscript on $F_\psi(x, B)$ by an $n$ when the argument $B$ is a subset of $\mathcal{X}^n$. Also, work with the product measure $\mathbb{Q} = \otimes_{i \leq n} Q_i$ for the distribution of $X$ and $\mathbb{Q}_{-n} = \otimes_{i < n} Q_i$ for the distribution of $(X_1, \ldots, X_{n-1})$. The assertion of the Theorem then becomes

$$\mathbb{Q} \exp\big(F_n(x, A)\big) \leq \exp(\gamma(\mathbb{Q}A))$$

For $n = 1$ and $B \in \mathcal{A}$ we have $F_1(x, B) = \psi(1)\{x \notin B\} + 0\{x \in B\}$ so that $Q_1 \exp(F_1(x, B)) \leq (1-p)e^{\psi(1)} + p$, where $p = Q_1 B$. Assumption (ii) then gives the desired $\exp(\gamma(p))$ bound.

Now suppose that $n > 1$ and that the inductive hypothesis is valid for dimensions strictly smaller than $n$. Write $\mathbb{Q}$ as $\mathbb{Q}_{-n} \otimes Q_n$. To simplify notation, write $w$ for $x_{-n} := (x_1, \ldots, x_{n-1})$ and $z$ for $x_n$. Define the cross section $A_z := \{w \in \mathcal{X}^{n-1} : (w, z) \in A\}$ and write $R_z$ for $\mathbb{Q}_{-n} A_z$. Define $r_0 := \sup_{z \in \mathcal{X}} R_z$. Notice that $r_0 \geq Q_n^z R_z = \mathbb{Q}A$.

The key to the proof is a recursive bound for $F_n$: for each $x = (w, z)$ with $A_z \neq \emptyset$, each $m$ with $A_m \neq \emptyset$, and all $\theta = 1 - \bar{\theta} \in [0, 1]$,

$$(7) \qquad \psi(\bar{\theta}) + F_n(x, A) \leq \theta F_{n-1}(w, A_z) + \bar{\theta} F_{n-1}(w, A_m).$$

To establish inequality (7), suppose $\mu_z$ is a probability measure concentrated on $A_z$ and $\mu_m$ is a probability measure concentrated on $A_m$. For a $\theta$ in $[0, 1]$, define $\nu = \theta \mu_z \otimes \delta_z + \bar{\theta} \mu_m \otimes \delta_m$, a probability measure concentrated on the subset $(A_z \times \{z\}) \cup (A_m \times \{m\})$ of $A$. Notice that, for $i < n$,

$$\nu\{y \in A : y_i \neq x_i\} = \theta \mu_z\{w \in A_z : y_i \neq x_i\} + \bar{\theta} \mu_m\{w \in A_m : y_i \neq x_i\}$$

so that, by convexity of $\psi$,

$$\psi\big(\nu\{y_i \neq x_i\}\big) \leq \theta \psi(\mu_z\{w \in A_z : y_i \neq x_i\}) + \bar{\theta}\psi(\mu_m\{w \in A_m : y_i \neq x_i\});$$

and (remembering that $x_n = z$),

$$\nu\{y \in A : y_n \neq x_n\} = \begin{cases} \bar{\theta} & \text{if } z \neq m \\ 0 & \text{otherwise} \end{cases} \leq \bar{\theta}.$$
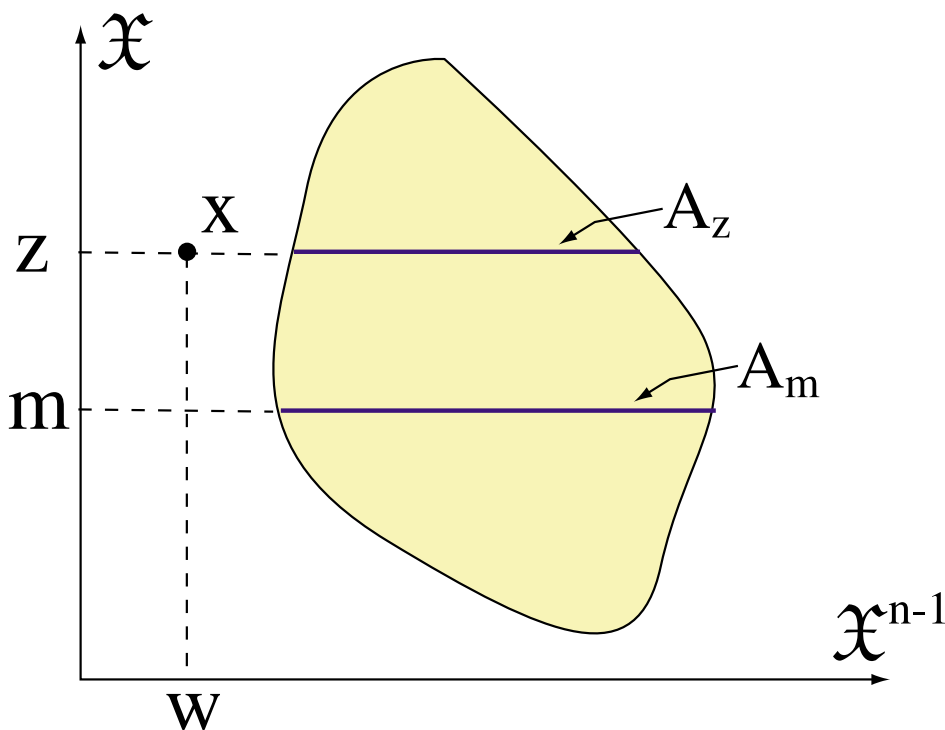
FIG 1.

Thus

$$F_n(x, A) \le \psi(\bar{\theta}) + \theta \sum_{i<n} \psi\left(\mu_z\{y_i \ne x_i\}\right) + \bar{\theta} \sum_{i<n} \psi\left(\mu_m\{y_i \ne x_i\}\right).$$

The two sums over the first $n-1$ coordinates are like those that appear in the definitions of $F_{n-1}(w, A_z)$ and $F_{n-1}(w, A_z)$. Indeed, taking an infimum over all $\mu_z \in \mathcal{P}(A_z)$ and $\mu_m \in \mathcal{P}(A_m)$ we get the expression on the right-hand side of (7).

Take exponentials of both sides of (7) then integrate out with respect to $\mathbb{Q}_{-n}$ over the $w$ component. For $0 < \theta < 1$ invoke the Hölder inquality, $\mathbb{Q}_{-n} U^\theta V^{\bar{\theta}} \le (\mathbb{Q}_{-n} U)^\theta (\mathbb{Q}_{-n} V)^{\bar{\theta}}$, with $U = \exp(F_{n-1}(w, A_z))$ and $V = \exp(F_{n-1}(w, A_m))$, for a fixed $m$. For each $z$ with $A_z \ne \emptyset$ we get

$$(8) \quad \begin{aligned} &\mathbb{Q}_{-n} \exp\left(F_n((w, z), A)\right) \\ &\qquad \le e^{\psi(\bar{\theta})} \left(\mathbb{Q}_{-n} \exp\left(F_{n-1}(w, A_z)\right)\right)^\theta \left(\mathbb{Q}_{-n} \exp\left(F_{n-1}(w, A_m)\right)\right)^{\bar{\theta}}. \end{aligned}$$

The inequality also hold in the extreme cases where $\theta = 0$ or $\theta = 1$, by continuity. The inductive hypothesis bounds the last product by

$$\exp\left(\psi(\bar{\theta}) + \theta \gamma(R_z) + \bar{\theta} \gamma(R_m)\right) = \exp\left(\gamma(R_m) + G(\gamma(R_z) - \gamma(R_m), \theta)\right).$$

The exponent is a decreasing function of $R_m$. Take an infimum over $m$, to replace $\gamma(R_m)$ by $\gamma(r_0)$. Then take an infimum over $\theta$ to get

$$(9) \quad \begin{aligned} &\mathbb{Q}_{-n} \exp\left(F_n((w, z), A)\right) \le \exp\left(\gamma(r_0) + G(\xi(R_z))\right) \\ &\qquad \text{where } \xi(r) := \gamma(R_z) - \gamma(r_0) \text{ for } 0 \le r \le r_0. \end{aligned}$$

If the cross section $A_z$ is empty, the set $\mathcal{P}(A_z)$ is empty. The argument leading from (7) to (9) still works if we fix $\theta$ equal to zero throughout, giving the bound

$$\mathbb{Q}^w_{-n} \exp\left(F_n(x, A)\right) \leq \exp\left(\gamma(r_0) + \psi(1)\right) \qquad \text{if } A_z = \emptyset.$$

Thus the inequality (9) also holds with $R_z = 0$ when $A_z = \emptyset$, because $\xi(0) = \gamma(0) - \gamma(r_0) = \infty$ and $G(\infty) = \psi(1)$.

By Assumption (i), the function $r \mapsto \exp(G(\xi(r)))$ is concave on $[0, r_0]$. Integrate both sides of (9) with respect to $Q_n$ to average out over the $z$ variable. Then invoke Jensen's inequality and the fact that $Q_n R_z = \mathbb{Q}A$, to deduce that

$$\mathbb{Q} \exp\left(F_n(x, A)\right) \leq \exp\left(\gamma(r_0) + G\left(\gamma(\mathbb{Q}A) - \gamma(r_0)\right)\right).$$

Finally, use the inequality $G(\eta) \leq \eta$ to bound the last expression by $\exp(\gamma(\mathbb{Q}A))$, thereby completing the inductive step.

**Remark.** Note that it is important to integrate with respect to $Q_n$ before using the bound on $G$: the upper bound $\exp(-\gamma(R_z))$ is a convex function of $R_z$, not concave.

## 3. Proof of the concavity lemma

I will establish a more detailed set of results than asserted by Lemma 2. Invoke the monotonicity and continuity of $\psi'$ to define $g(\eta)$ as the solution to $\psi'\left(1 - g(\eta)\right) = \eta$ if $0 \leq \eta < \psi'(1)$ and $g(\eta) = 0$ if $\psi'(1) \leq \eta$. Then the following assertions are true. (I drop the $\psi$ subscripts for notational simplicity.)

(i)
$$G(\eta) = \begin{cases} \psi\left(1 - g(\eta)\right) + \eta g(\eta) & \text{for } 0 \leq \eta < \psi'(1) \\ \psi(1) & \text{for } \psi'(1) \leq \eta \end{cases}$$

(ii) $G$ is increasing and concave, with a continuous, decreasing first derivative $g$. In particular, $G(0) = 0$ and $G'(0) = g(0) = 1$.

(iii) $G''(\eta) = g'(\eta) = -[\psi''(1 - g(\eta))]^{-1}$ for $0 < \eta < \psi'(1)$.

(iv) $G(\eta) \leq \eta$ for all $\eta \in \mathbb{R}^+$.

(v) Suppose $\xi : J \to \mathbb{R}^+$ is a convex function defined on a subinterval $J$ of the real line, with $\xi' \neq 0$ on the interior of $J$. Suppose

$$\frac{1}{\psi''(1 - \xi_r)} \geq g(\xi_r)^2 + g(\xi_r)\xi''(r)/\xi'(r)^2,$$

for all $r$ in the interior of $J$ for which $\xi_r := \xi(r) \in (0, 1)$. Then $r \mapsto \exp(G(\xi(r)))$ is a concave function on $J$.

*Proof of (i) through (iv).* The fact that $G$ is concave and increasing follows from its definition as an infimum of increasing linear functions of $\eta$. (It would also follow from the fact that $G'(\eta) = g(\eta)$, which is nonnegative and decreasing.) Replacement of the infimum over $0 \leq \theta \leq 1$ by the value at $\theta = 1$ gives the inequality $G(\eta) \leq \eta$.

If $\eta \geq \psi'(1)$, the derivative $-\psi'(1 - \theta) + \eta$ is nonnegative on $[0, 1]$, which ensures that the infimum is achieved at $\theta = 1$.

If $0 < \eta < \psi'(1)$, the infimum is achieved at the zero of the derivative, $\theta = g(\eta)$. Differentiation of the defining equality $\psi'(1 - g(\eta)) = \eta$ then gives the expression for $g'(\eta)$. Similarly

$$G'(\eta) = -\psi'\left(1 - g(\eta)\right) g'(\eta) + \eta g'(\eta) + g(\eta) = g(\eta).$$

The infimum that defines $G(0)$ is achieved at $g(0) = 1$, which gives $G(0) = \psi(0) = 0$. Continuity of $g$ at then gives $G'(0) = g(0) = 1$. □

*Proof of (v).* Note that the function $L(r) := \exp(G(\xi(r)))$ is continuous on $J$ and takes the value $e^{\psi(1)}$ for all $r$ at which $\xi(r) \geq \psi'(1)$. The second derivative $L''(r)$ exists except possibly at points $r$ for which $\xi(r) = \psi'(1)$. In particular, $L''(r) = 0$ when $\xi(r) > \psi'(1)$ and

$$L''(r) = \left( g'(\xi_r)(\xi_r')^2 + g(\xi_r)\xi_r'' + g(\xi_r)^2(\xi_r')^2 \right) L(r) \qquad \text{for } 0 < \xi_r < \psi'(1).$$

From (iii) and the positivity of $L$, the last expression is $\leq 0$ if and only if

$$-\frac{(\xi_r')^2}{\psi''(1 - g(\xi_r))} + g(\xi_r)\xi_r'' + g(\xi_r)^2(\xi_r')^2 \leq 0.$$

Divide through by $(\xi_r')^2$ then rearrange to get the asserted inequality for $\psi''$. Lemma 2 follows as a special case of (i) through (iv). □

**Special cases.** If $\sup_r \xi''(r)/\xi'(r)^2 \leq c$, with $c$ a positive constant, the inequality from part (v) will certainly hold if

$$(10) \qquad\qquad \psi''(1 - \theta) \leq (\theta^2 + c\theta)^{-1} \qquad \text{for all } 0 < \theta < 1.$$

This differential inequality can be solved, subject to the constraints $0 = \psi(0) = \psi'(0)$, by two integrations. Indeed,

$$\psi'(1 - \theta) = \int_\theta^1 \psi''(1 - t)\, dt \leq \int_\theta^1 \frac{dt}{t^2 + ct} = c^{-1}\left( -\log\theta + \log\left(\frac{\theta + c}{1 + c}\right) \right)$$

and, with $\psi_c$ defined by (6),

$$\psi(1 - \theta) = \int_\theta^1 \psi'(1 - t)\, dt \leq c^{-1}\int_\theta^1 -\log t + \log\left(\frac{t + c}{1 + c}\right) dt = \psi_c(1 - \theta).$$

Note that $\psi_c(1 - \theta)$ is the solution to the differential equation

$$\psi_c''(1 - \theta) = \frac{1}{\theta^2 + c\theta} \qquad \text{for all } 0 < \theta < 1, \text{ with } \psi_c(0) = \psi_c'(0) = 0.$$

It is the largest solution to (10).

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