

Generating families of continuous univariate distributions

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- Koutras, M. V. and Dafnis, S. (2024). A new family of continuous univariate distributions. Under revision in *Methodology and Computing in Applied Probability*.
- Koutras, M. V. and Dafnis, S. (2024). On the generation of continuous univariate distributions. Submitted for publication.

Generating new families of distributions from existing ones

FAMILY 1

Azzalini-Type
Skew-Symmetric

FAMILY 2

Transformation of
Random Variable

FAMILY 3

Transformation of
Scale

FAMILY 4

Probability Integral
Transformation of a
Random Variable
on $[0,1]$

Generating new families of distributions from existing ones

FAMILY 1

Azzalini-Type
Skew-Symmetric

Define the density of X_A to be

$$f_A(x) = 2W(x)f(x)$$

where f is a pdf and $W(\cdot)$ a function such that

$$W(x) + W(-x) = 1$$

The most familiar special cases take $W(x) = F(\alpha x)$ to be the cdf of a (scaled) symmetric distribution

Azzalini, 1985, *Scand. J. Stat.*, Azzalini with Capitanio, 2014, *book*

Generating new families of distributions from existing ones

FAMILY 2

Transformation of Random Variable

Let $W: R \rightarrow R$ be an invertible increasing function. If $X \sim f$, then define $X_T = W(X)$. The density of the distribution of X_T is

$$f_T(x) = \frac{f(W^{-1}(x))}{W'(W^{-1}(x))}$$

Jones & Pewsey, 2009, *Biometrika*

Generating new families of distributions from existing ones

FAMILY 3

Transformation of
Scale

The density of the distribution of X_S is just

$$f_S(x) = 2f(W^{-1}(x))$$

which is a density if

$$W(x) - W(-x) = x$$

Jones, 2014, *Statist. Sinica*

Generating new families of distributions from existing ones

FAMILY 4

Probability Integral
Transformation of a
Random Variable
on $[0,1]$

Let g be the density of a random variable U in $(0,1)$. Then define $X_U = F^{-1}(U)$ where $F' = f$. The density of the distribution of X_U is

$$f_U(x) = f(x)g(F(x))$$

An easy task, but

It is **one of the easiest things in statistics** to invent new univariate distributions; after all, any non-negative integrable function is the core of a density function. The ongoing challenge is to extract from the overwhelming plethora of possibilities those relatively few with the **best and most appropriate properties that are of real potential value in practical applications.**

Jones, M. C. (2015). On families of distributions with shape parameters. *International Statistical Review* 83, 2, 175-192.

Motivation

For many classical continuous univariate distributions, there exists a monotone transformation $g(\cdot)$, of their cumulative density function (cdf) $F(x)$, so that

$$g(F(x)) = h(x; \theta),$$

where

- $g(x)$ does not involve any of the distributions parameters.
- $h(x; \theta)$ contains the parameters of the distribution.

Well known distributions

$$g(F(x)) = h(x; \theta)$$

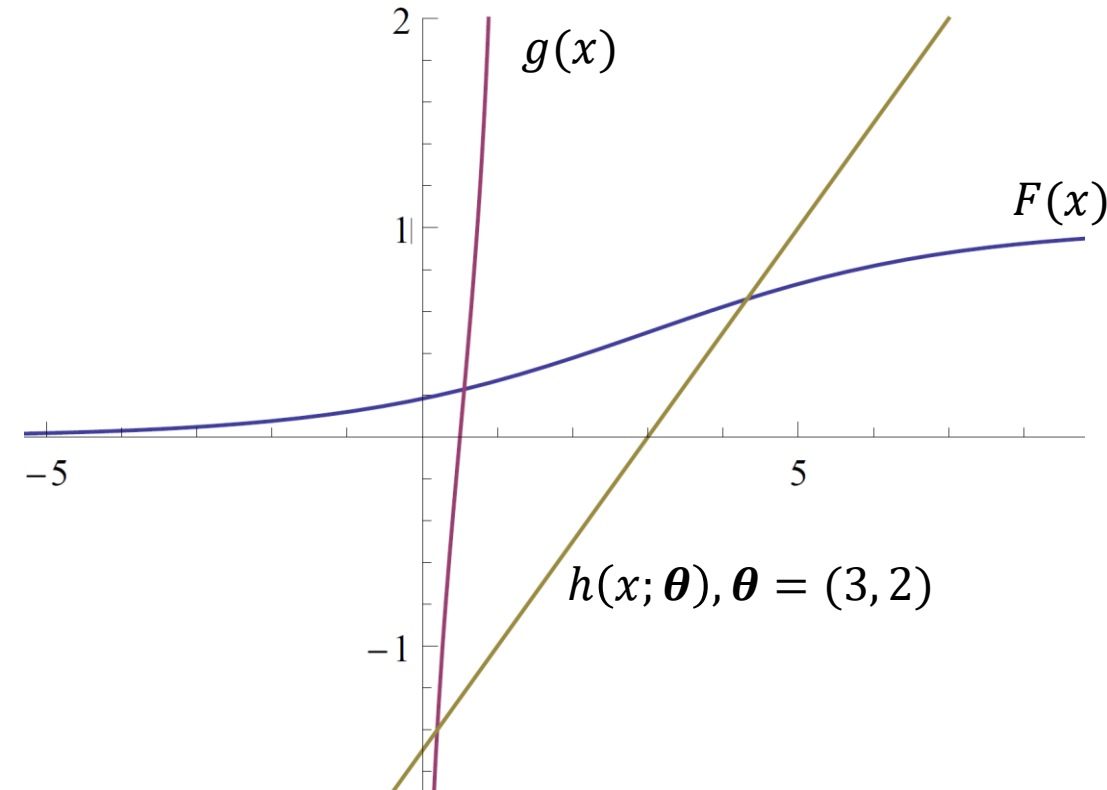
Distribution	$F(x)$	θ	Transformation	$g(x)$	$h(x; \theta)$
Exponential	$1 - e^{-\lambda x}$	λ	$-\ln(1 - F) = \lambda x$	$-\ln(1 - x)$	λx
Weibull	$1 - e^{-\lambda x^\alpha}$	(α, λ)	$-\ln(1 - F) = \lambda x^\alpha$	$-\ln(1 - x)$	λx^α
Pareto known α	$1 - (\frac{\lambda}{x+\lambda})^\alpha$	λ	$(\frac{1}{1-F})^{\frac{1}{\alpha}} = \frac{x}{\lambda} + 1$	$(\frac{1}{1-x})^{\frac{1}{\alpha}}$	$\frac{x}{\lambda} + 1$
Gompertz	$1 - e^{-\alpha(e^{\lambda x} - 1)}$	(α, λ)	$-\ln(1 - F) = \alpha(e^{\lambda x} - 1)$	$-\ln(1 - x)$	$\alpha(e^{\lambda x} - 1)$
Dagum known p	$(1 + (\frac{x}{\lambda})^{-\alpha})^{-p}$	(α, λ)	$(F^{-1/p} - 1)^{-1} = (\frac{x}{\lambda})^\alpha$	$\frac{1}{-1+x^{-1/p}}$	$(\frac{x}{\lambda})^\alpha$
Exponential-logarithmic known p	$1 - \frac{\ln(1-(1-p)e^{-\lambda x})}{\ln p}$	λ	$-\ln(\frac{1-p^{1-F}}{1-p}) = \lambda x$	$-\ln(\frac{1-p^{1-x}}{1-p})$	λx
Log-logistic	$\frac{x^\alpha}{\lambda^\alpha + x^\alpha}$	α	$\ln(\frac{F}{1-F}) = \alpha \ln(\frac{x}{\lambda})$	$\ln(\frac{x}{1-x})$	$\alpha \ln(\frac{x}{\lambda})$
Burr known α	$1 - (\frac{1}{1+(\frac{x}{\lambda})^\beta})^\alpha$	β, λ	$(\frac{1}{1-F})^{\frac{1}{\alpha}} = (\frac{x}{\lambda})^\beta + 1$	$(\frac{1}{1-x})^{\frac{1}{\alpha}}$	$(\frac{x}{\lambda})^\beta + 1$

Example: Logistic Distribution

$$g(F(x)) = h(x; \theta)$$

$$F(x) = \frac{1}{1 + e^{-\frac{x-\mu}{\sigma}}}, \quad x \in \mathcal{R}$$

- Transformation: $\ln\left(\frac{F}{1-F}\right) = \frac{x-\mu}{\sigma}$
- Generator: $g(x) = \ln\left(\frac{x}{1-x}\right)$
- Parametric part: $h(x; \theta) = \frac{x-\mu}{\sigma}$



The New Family: Definition

$$g(F(x)) = h(x; \theta)$$

We shall say that a distribution with support $(-\infty, \infty)$ and cdf F belongs to the class $D_g(h)$ (notation: $F \in D_g(h)$) if

$$F(x) = g^{-1}(h(x))$$

where

- C1.** $g: (0,1) \rightarrow \mathcal{R}$ is strictly increasing.
- C2.** $h: (-\infty, \infty) \rightarrow \mathcal{R}$ is increasing.
- C3.** g, h differentiable.
- C4.** $\lim_{x \rightarrow 1^-} g(x) = \infty$ and $\lim_{x \rightarrow 0^+} g(x) = l$ ($l \in \mathcal{R}$ or $l = -\infty$).
- C5.** $\lim_{x \rightarrow \infty} h(x) = \infty$ and $\lim_{x \rightarrow -\infty} h(x) = l$.

Aging Properties of $D_g(h)$ – Proposition 1a

Let $F \in D_g(h)$, $\Delta \subseteq \mathcal{R}$ and $Q(\cdot)$ the function defined by

$$Q(x) = g'(x)(1 - x), \quad 0 < x < 1.$$

If

- h' is decreasing in Δ (h concave) and
- Q is increasing in $F(\Delta) \subseteq (0, 1)$

then F has a decreasing failure rate ($F \in DFR$) in Δ .

.... best and most appropriate properties that are of real potential value in practical applications

The failure rate of $D_g(h)$ – sketch of Proposition's proof

Let $F \in D_g(h)$, $\Delta \subseteq \mathcal{R}$ and $Q(\cdot)$ the function defined by

$$Q(x) = g'(x)(1 - x), \quad 0 < x < 1.$$

If

- h' is decreasing in Δ (h concave) and
- Q is increasing in $F(\Delta) \subseteq (0, 1)$

then F has a decreasing failure rate ($F \in DFR$) in Δ .

$$r(x) = \frac{h'(x) \downarrow}{Q \uparrow (F \uparrow (x))}$$

Aging Properties of $D_g(h)$ – Proposition 1b

Let $\Delta \subseteq \mathcal{R}$. If

- h' is decreasing in Δ (h concave) and
- Q is increasing in $F(\Delta) \subseteq (0, 1)$

then $F \in D_g(h)$ has a decreasing failure rate ($F \in DFR$) in Δ .

Let $\Delta \subseteq \mathcal{R}$. If

- h' is increasing in Δ (h convex) and
- Q is decreasing in $F(\Delta) \subseteq (0, 1)$

then $F \in D_g(h)$ has an increasing failure rate ($F \in IFR$) in Δ .

Aging Properties of $D_g(h)$ – Proposition 1c

Let $\Delta \subseteq \mathcal{R}$ and assume that h' is constant in Δ (i.e. the parametric part h is linear in x in Δ).

- If Q increasing in $F(\Delta) \subseteq (0, 1)$, then $F \in D_g(h)$ is *DFR* in Δ .
- If Q decreasing in $F(\Delta) \subseteq (0, 1)$, then $F \in D_g(h)$ is *IFR* in Δ .
- If Q constant in $F(\Delta) \subseteq (0, 1)$, then $F \in D_g(h)$ has a constant failure rate in Δ .

$$r(x) = \frac{h'(x)}{Q(F(x))} = \frac{\text{constant}}{Q(F(x))}$$

Aging Properties of classical distributions exploiting Proposition 1c

For many classical distributions we have

$$h(x) = cx + d, \quad c > 0.$$

Therefore, the study for the aging properties of the classical distributions can be conferred from the monotonicity properties of the function

$$Q(x) = g'(x)(1 - x), \quad 0 < x < 1.$$

Aging Properties of classical distributions exploiting Proposition 1c

	Distribution	$F(x)$	$g(x)$	$h(x; \theta)$
1	Logistic	$\frac{1}{1 + e^{-\frac{x-\mu}{\sigma}}}$	$\ln\left(\frac{x}{1-x}\right)$	$\frac{x-\mu}{\sigma}$
2	Gumbel	$\exp\left(-e^{-\frac{x-\mu}{\sigma}}\right)$	$-\ln(-\ln x)$	$\frac{x-\mu}{\sigma}$
3	Cauchy	$\frac{1}{2} + \frac{1}{\pi} \arctan\left(-\frac{x-\mu}{\sigma}\right)$	$\tan\left(\pi x - \frac{\pi}{2}\right)$	$\frac{x-\mu}{\sigma}$

$$Q(x) = g'(x)(1-x)$$

Proposition 1c: Application for the Logistic Distribution

$$g(x) = \ln\left(\frac{x}{1-x}\right), \quad h(x; \boldsymbol{\theta}) = \frac{x-\mu}{\sigma}$$

$$Q(x) = g'(x)(1-x) = \frac{1}{x}$$

$$Q'(x) = -\frac{1}{x^2} < 0$$

Q is decreasing in $F((-\infty, \infty)) = (0, 1)$

$\rightarrow F$ is *IFR* in $(-\infty, \infty)$

Let $\Delta \subseteq \mathcal{R}$ and assume that h' is constant in Δ .

- If Q increasing in $F(\Delta) \subseteq (0, 1)$, then $F \in IFR$ in Δ .
- If Q decreasing in $F(\Delta) \subseteq (0, 1)$, then $F \in IFR$ in Δ .
- If Q constant in $F(\Delta) \subseteq (0, 1)$, then F has a constant failure rate in Δ .

Probability Bounds for $D_g(h)$

Proposition 3. If $F \in D_g(h)$ and c is a positive lower bound for the function g' , i.e.

$$g'(x) \geq c, \quad \forall x \in (0,1)$$

then

$$P(x_1 < X < x_2) \leq \frac{h(x_2; \boldsymbol{\theta}) - h(x_1; \boldsymbol{\theta})}{c},$$

for every x_1, x_2 with $x_1 < x_2$.

Unimodality of $D_g(h)$

Proposition 2. Let $F \in D_g(h)$ and $w(\cdot)$ the function defined in $\Delta \subseteq \mathcal{R}$ by the formula

$$w(x) = \frac{1}{g'(x)}.$$

.... best and most appropriate properties that are of real potential value in practical applications

If

- w is concave in $F(\Delta)$
 - h' is logconcave in Δ
 - both h and g are concave functions or both h and g are convex functions in Δ and $F(\Delta) \subseteq (0, 1)$, respectively,
- then $F \in D_g(h)$ is unimodal and *IFR* in Δ .

A gallery for h functions to generate distributions in $D_g(h)$

➤ Linear

$$h(x) = \alpha x, \quad \alpha > 0$$

➤ Power

$$h(x) = \alpha x^b, \quad \alpha > 0 \text{ and } b \geq 1 \text{ odd}$$

➤ Exponential

$$h(x) = a e^{bx} \quad \alpha, b > 0$$

➤ Exponential – Logarithmic

$$h(x) = \ln(\ln(a + b e^{cx})), \quad b, c > 0 \text{ and } a \geq 1$$

➤ Power - Logarithmic

$$h(x) = a \ln(1 + b x^c) \quad a > 0, b > 0 \text{ and } c \geq 1 \text{ odd}$$

C2. $h: (-\infty, \infty) \rightarrow \mathcal{R}$ is increasing.

C3. h differentiable.

C5. $\lim_{x \rightarrow \infty} h(x) = \infty$ and $\lim_{x \rightarrow -\infty} h(x) = l$
($l \in \mathcal{R}$ or $l = -\infty$).

Generation of new distributions in $D_g(h)$: linear combinations

Proposition 4. Let

$$F_1 \in D_g(h_1), \quad F_2 \in D_g(h_2)$$

and introduce the function h_3 defined by

$$h_3 = b_1 h_1 + b_2 h_2,$$

with $b_1, b_2 > 0$. Then $F_3 = g^{-1} \circ h_3 \in D_g(h_3)$ if

- $\lim_{x \rightarrow 0^+} g(x) = l \in \mathcal{R}$ and $b_1 + b_2 = 1$ or
- $\lim_{x \rightarrow 0^+} g(x) = -\infty$ or
- $\lim_{x \rightarrow 0^+} g(x) = 0$.

Generation of new distributions in $D_g(h)$: multiplication

Proposition 5. Let

$$F_1 \in D_g(h_1), \quad F_2 \in D_g(h_2)$$

and introduce the function h_3 defined by

$$h_3 = h_1 h_2.$$

If $\lim_{x \rightarrow 0^+} g(x) = l \in \mathcal{R}$ and $l = 0$ or $l = 1$, then

$$F_3 = g^{-1} \circ h_3 \in D_g(h_3).$$

Generation of new distributions in $D_g(h)$: composition

Proposition 6. Let

$$F_1 \in D_g(h_1), \quad F_2 \in D_g(h_2)$$

and introduce the function h_3 defined by

$$h_3 = h_1 \circ h_2.$$

If

$$\lim_{x \rightarrow 0^+} g(x) = l \in \mathcal{R} \text{ and } h_1(l) = l,$$

then

$$F_3 = g^{-1} \circ h_3 \in D_g(h_3).$$

Generation of new distributions in $D_g(h)$:transforming the generator

Proposition 4.1. *Let $F \in D_g^+(h)$ and $g_0 : (0, 1) \rightarrow \mathbb{R}$ a strictly increasing and differentiable function such that*

$$\lim_{x \rightarrow 0^+} g_0(x) = 0 \text{ and } \lim_{x \rightarrow 1^-} g_0(x) = 1.$$

Then, the function $g_1 = g \circ g_0$ is a valid generator leading to the cdf

$$F^* = g_1^{-1} \circ h \in D_{g_1}^+(h).$$

Generation of new distributions in $D_g(h)$:transforming the generator

	$g_0(x)$	$g_0^{-1}(x)$	Parameters	Comments
1	$(-\ln(1 - x(1 - e^{-1})))^{\frac{1}{\alpha}}$	$\frac{e^{-x^\alpha} - 1}{e^{-1} - 1}$	$\alpha > 0$	Khalil et al. (2021)
2	$\frac{(1-\theta)x}{x-\theta}$	$\frac{\theta x}{\theta-1+x}$	$\theta > 0$	Ahmad et al. (2022)
3	$\frac{\ln((\theta-1)x+1)}{\ln\theta}$	$\frac{\theta^x - 1}{\theta - 1}$	$0 < \theta \neq 1$	Mahdavi and Kundu (2017)
4	$\left(1 - \frac{\theta}{\ln(1-x^{\frac{1}{\alpha}})}\right)^{-1}$	$(1 - e^{-\theta \frac{x}{1-x}})^\alpha$	$\theta > 0, \alpha > 0$	Barati and Rashidi (2022)
5	x^α	$x^{\frac{1}{\alpha}}$	$\alpha > 0$	Special case of the $Beta(\alpha, \beta)$ distribution with $\beta = 1$
6	$1 - (1 - x)^\beta$	$1 - (1 - x)^{\frac{1}{\beta}}$	$\beta > 0$	Special case of the $Beta(\alpha, \beta)$ distribution with $\alpha = 1$
7	$1 - (1 - x^\alpha)^\beta$	$(1 - (1 - x)^{\frac{1}{\beta}})^{\frac{1}{\alpha}}$	$\alpha, \beta > 0$	Kumaraswamy Distribution, see e.g. Jones (2009)
8	$\frac{1}{1-\ln x}$	$e^{1-\frac{1}{x}}$	-	

Properties of the transformed family

Proposition 5.4. *Let $F \in D_g^+(h)$, $g_0 : (0, 1) \rightarrow \mathbb{R}$ a transformation function and $F^* \in D_{g,g_0}^+(h)$ the transformed D_g^+ -family. Let also $s(x)$ denote the function*

$$s(x) = \frac{g_0'(x)(1-x)}{1-g_0(x)} = -(1-x)(\ln(1-g_0(x)))', \quad 0 < x < 1.$$

- a. *If F is IFR on $\Delta \subseteq (0, \infty)$ and $s(x)$ is decreasing on $F(\Delta)$ then F^* is IFR on Δ .*
- b. *If F is DFR on $\Delta \subseteq (0, \infty)$ and $s(x)$ is increasing on $F(\Delta)$, then F^* is DFR on Δ .*

Properties of the transformed family

Proposition 5.8. *Let $F \in D_g^+(h)$, $g_0 : (0, 1) \rightarrow \mathbb{R}$ a transformation function with*

$$\lim_{x \rightarrow 1^-} g_0'(x) \neq 0$$

and $F^ \in D_{g, g_0}^+(h)$ the corresponding transformed D_g^+ – family.*

- a. If F is heavy tailed, then F^* is heavy tailed.*
- b. If F is not heavy tailed, then F^* is not heavy tailed.*

Further research

- Additional closure properties (when combining several parametric parts)
- Study of fitting performance in
 - experimental data (hydration heat, antibacterial activity)
 - collections of big data (social networking data)

Thank you for your attention!