

# A formulation for continuous mixtures of multivariate normal distributions

Adelchi Azzalini

Università di Padova, Italia

joint work with [Reinaldo B. Arellano-Valle](#)

Pontificia Universidad Católica de Chile

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# Continuous mixtures of multivariate normal variables

Start from  $X \sim N_d(0, \Sigma)$ ,  $\Sigma > 0$ , and let  $\xi \in \mathbb{R}^d$  be a fixed vector.

Then combine  $X$  with another (scalar, indept) random variable, such as:

- 1  $\sqrt{V} X$  for some random  $V > 0 \implies$  *scale mixture* of  $X$   
(a wide subset of the elliptical class of distributions);
- 2  $U \xi + X$  for some random  $U \implies$  *mean mixture*;
- 3  $V \xi + \sqrt{V} X \implies$  *variance-mean mixture*  
(if  $V \sim$  GIG, get the Generalized Hyperbolic distribution).

Considerable literature exists, especially for the forms 1 and 3.  
(impossible to review it here; see the paper for references)

In recent years, mixtures of other distributions have been considered, especially with  $X \sim$  SkewNormal.

## A two-component formulation

- Re-start from  $X \sim N_d(0, \Sigma)$ , but now use *two* mixing terms
- Let  $U, V$  two scalar variables, such that  $X, U, V$  are all independent.
- Given a real-valued function  $r$ , a positive-valued function  $s$  and  $\gamma \in \mathbb{R}^d$ , we consider

$$\begin{aligned} Y &= \xi + r(U, V)\gamma + s(U, V)X \\ &= \xi + R\gamma + SX \end{aligned}$$

called a *generalized mixture of normals* (GMNs).

# Aims

- The basic aim is to highlight a common framework:  
many existing mixtures (not only of normals) are GMNs
- Hence obtain a better understanding of their nature and connections
- Also, can we carry out a unified treatment of the properties?
- Finally, the GMNs scheme might suggest new constructions

# First general implications

$$Y = \xi + R\gamma + SX$$

- The relevant ingredients are  $\xi, \gamma, \Sigma$  and the distribution of  $(R, S) \sim H$
- So write  $Y \sim \text{GMN}_d(\xi, \Sigma, \gamma, H)$
- An affine transformation of  $Y$  is still of type GMN
- Hence the GMN class is closed under marginalization
- $\mathbb{E}\{Y\} = \xi + \mathbb{E}\{R\}\gamma$ , if the moment exists
- $\text{var}\{Y\} = \text{var}\{R\}\gamma\gamma^\top + \mathbb{E}\{S^2\}\Sigma$ , if the moments exist

## Results after some algebra (... sometimes a lot of it)

- If  $Y$  is partitioned into  $Y_1$  and  $Y_2$ , the conditional distribution of  $Y_1$  given  $Y_2$ , which is still of GMN type
- A collection of facts about quadratic forms of  $Y$
- An expression for the Mardia's measures of multivariate asymmetry and kurtosis, involving moments of  $(R, S)$  up to the fourth order

# Classical mixtures of normals

- $(R, S) = (0, S) \implies$  scale/variance mixtures of normals
- $(R, S) = (R, 1) \implies$  mean mixtures of normals
- $(R, S) = (V, \sqrt{V})$  with  $V > 0 \implies$  variance-mean mixtures

Reproduce a wide range of classical constructions

## Aside: the skew-normal (SN) distribution – 1

$$\text{SN.density}(x) = 2 \text{Normal.density}(x) \Phi(\eta^\top x)$$

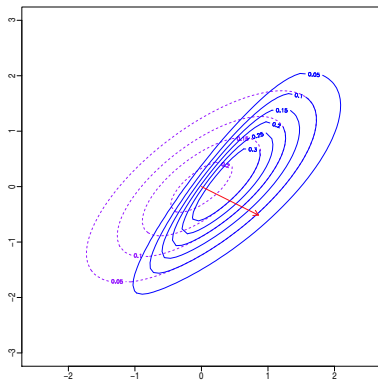
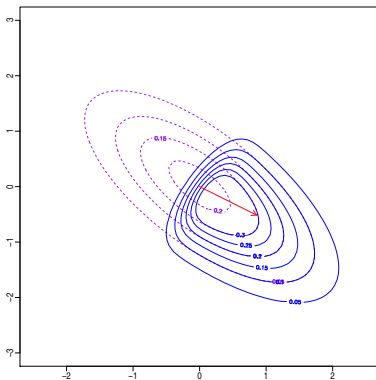
- Extends the Normal distribution with extra parameter  $\eta \in \mathbb{R}^d$
- If  $\eta = 0$  back to Normal, otherwise the density is asymmetric
- Numerous formal properties are available
- Representation as **mean mixture** of Normals:

$$\xi + R\gamma + X \equiv \xi + U\gamma + X$$

where  $R \equiv U \sim \chi_1$  and  $\gamma$  is related to  $\eta$ .



## Aside: the skew-normal (SN) distribution – 2



# Mixtures of SN distributions

- If  $Z \sim \text{SN}$ , its scale mixture takes the form, dropping  $\xi$ ,

$$\sqrt{V} Z = \sqrt{V} (U\gamma + X) = U \sqrt{V}\gamma + \sqrt{V} X$$

hence this is a GMN with  $R = \sqrt{V} U$ ,  $S = \sqrt{V}$ .

Can regulate both skewness and tail weight

(Example: a popular instance is the skew- $t$  distribution)

- Scale mixing can be combined with shape mixing

$$U\gamma \implies U(\tilde{U}\tilde{\gamma}), \quad \sqrt{V}U\gamma \implies \sqrt{V}(U\tilde{U})\tilde{\gamma}$$

hence GMN with  $R = \sqrt{V}(U\tilde{U}) = \sqrt{V}U^*$  and  $S = \sqrt{V}$ .

# An exploration with mean mixtures

$$Y = \xi + R\gamma + Z$$

- If  $R \sim \chi_1$ , get the SN distribution
- Another option: take  $R \sim \chi_2$  (Rayleigh distribution)
- More generally,  $R \sim \chi_\nu$  (Nakagami  $m$ -distribution)

# Final comments

- The GMN class includes many forms of mixtures, not all
- Extensions are possible in a number of directions
- However, these will come at some cost, like loss of some formal properties, technical difficulties, etc.
- GMN seeks a balance between generality and tractability

# The end

The full story:

R. B. Arellano-Valle and A. Azzalini

A formulation for continuous mixtures of multivariate normal distributions.

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Thanks for your attention.