## Sum of Two Standard Uniform Random Variables



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Based on joint work with Bin Wang (Beijing)

## A Question

In this talk we discuss this problem:

$$
\begin{aligned}
& \quad X_{1} \sim \mathrm{U}[-1,1], X_{2} \sim \mathrm{U}[-1,1] \\
& \text { what is a distribution }(\mathrm{cdf}) \text { of } X_{1}+X_{2} \text { ? }
\end{aligned}
$$




A difficult problem with no applications (?)

## Generic Formulation

In an atomless probability space:

- $F_{1}, \ldots, F_{n}$ are $n$ distributions
- $X_{i} \sim F_{i}, i=1, \ldots, n$
- $S_{n}=X_{1}+\cdots+X_{n}$

Denote the set of possible aggregate distributions

$$
\mathcal{D}_{n}=\mathcal{D}_{n}\left(F_{1}, \cdots, F_{n}\right)=\left\{\text { cdf of } S_{n} \mid X_{i} \sim F_{i}, i=1, \cdots, n\right\} .
$$

Primary question: Characterization of $\mathcal{D}_{n}$.

- $\mathcal{D}_{n}$ is non-empty, convex, and closed w.r.t. weak convergence


## Generic Formulation

For example:

- $X_{i}$ : individual risks; $S_{n}$ : risk aggregation
- fixed marginal; unknown copula

Classic setup in Quantitative Risk Management

- Secondary question: what is $\sup _{F \in \mathcal{D}_{n}} \rho(F)$ for some functional $\rho$ (risk measure, utility, moments, ...)?
- Risk aggregation with dependence uncertainty, an active field over the past few years:
- Embrechts et. al. (2014 Risks) and the references therein
- Books: Rüschendorf (2013), McNeil-Frey-Embrechts (2015)
- $20+$ papers in the past 3 years


## Some Observations

Assume that $F_{1}, \ldots, F_{n}$ have finite means $\mu_{1}, \ldots, \mu_{n}$, respectively.

- Necessary conditions:
- $S_{n} \prec_{\mathrm{cx}} F_{1}^{-1}(U)+\cdots+F_{n}^{-1}(U)$
- In particular, $\mathbb{E}\left[S_{n}\right]=\mu_{1}+\cdots+\mu_{n}$
- Range $\left(S_{n}\right) \subset \sum_{i=1}^{n} \operatorname{Range}\left(X_{i}\right)$
- Suppose $\mathbb{E}[T]=\mu_{1}+\cdots+\mu_{n}$. Then

$$
F_{T} \in \mathcal{D}_{n}\left(F_{1}, \ldots, F_{n}\right) \Leftrightarrow\left(F_{-T}, F_{1}, \ldots, F_{n}\right) \text { is jointly mixable }
$$

For a theory of joint mixability

- W.-Peng-Yang (2013 FS), Wang-W. (2016 MOR)
- Surveys: Puccetti-W. (2015 STS), W. (2015 PS)
- Numerical method: Puccetti-W. (2015 JCAM)


## Some Observations

- Joint mixability is an open research area
- A general analytical characterization of $\mathcal{D}_{n}$ or joint mixability is far away from being clear
- We tune down and look at standard uniform distributions and $n=2$


## Progress of the Talk


(2) Some Examples
(3) Some Answers
(4) Some More
(5) References

## Simple Examples

$$
X_{1} \sim \mathrm{U}[-1,1], X_{2} \sim \mathrm{U}[-1,1], S_{2}=X_{1}+X_{2} .
$$




Obvious constraints

- $\mathbb{E}\left[S_{2}\right]=0$
- range of $S_{2}$ in $[-2,2]$
- $\operatorname{Var}\left(S_{2}\right) \leq 4 / 3$
- $S_{2} \prec_{\mathrm{cx}} 2 X_{1}$
(sufficient?)


## Simple Examples

Are the following distributions possible for $S_{2}$ ?







## Simple Examples: More...







Examples and counter-examples: Mao-W. (2015 JMVA) and Wang-W. (2016 MOR)

## A Small Copula Game...

$$
\mathbb{P}\left(S_{2}=-4 / 5\right)=1 / 2, \mathbb{P}\left(S_{2}=4 / 5\right)=1 / 2
$$



## Progress of the Talk

## (1) Question

(2) Some Examples
(3) Some Answers

4 Some More
(5) References

## Existing Results

Let $\mathcal{D}_{2}=\mathcal{D}_{2}(\mathrm{U}[-1,1], \mathrm{U}[-1,1])$. Below are implied by results in
Wang-W. (2016 MOR)

- Let $F$ be any distribution with a monotone density function. then $F \in \mathcal{D}_{2}$ if and only if $F$ is supported in $[-2,2]$ and has zero mean.
- Let $F$ be any distribution with a unimodal and symmetric density function. Then $F \in \mathcal{D}_{2}$ if and only if $F$ is supported in $[-2,2]$ and has zero mean.
- $\mathrm{U}[-a, a] \in \mathcal{D}_{2}$ if and only if $a \in[0,2]$ (a special case of both).
- The case $\mathrm{U}[-1,1] \in \mathcal{D}_{2}$ is given in Rüschendorf (1982 JAP).


## Unimodal Densities

A natural candidate to investigate is the class of distributions with a unimodal density.

## Theorem 1

Let $F$ be a distribution with a unimodal density on $[-2,2]$ and zero mean. Then $F \in \mathcal{D}_{2}$.

- Both the two previous results are special cases
- For bimodal densities we do not have anything concrete


## Densities Dominating a Uniform

A second candidate is a distribution which dominates a portion of a uniform distribution.

## Theorem 2

Let $F$ be a distribution supported in $[a-b, a]$ with zero mean and density function $f$. If there exists $h>0$ such that $f \geq \frac{3 b}{4 h}$ on [ $-h / 2, h / 2$ ], then $F \in \mathcal{D}_{2}$.

- The density of $F$ dominates $3 b / 4$ times that of $\mathrm{U}[-h / 2, h / 2]$


## Bi-atomic Distributions

Continuous distributions seem to be a dead end; what about discrete distributions? Let us start with the simplest cases.

## Bi-atomic Distributions

## Theorem 3

Let $F$ be a bi-atomic distribution with zero mean supported on $\{a-b, a\}$. Then $F \in \mathcal{D}_{2}$ if and only if $2 / b \in \mathbb{N}$.

$2 / b=1$

$2 / b=5 / 4$

- For given $b>a>0$, there is only one distribution on $\{a-b, a\}$ with mean zero.


## Tri-atomic Distributions

For a tri-atomic distribution $F$, write $F=\left(f_{1}, f_{2}, f_{3}\right)$ where $f_{1}, f_{2}, f_{3}$ are the probability masses of $F$

- On given three points, the set of tri-atomic distributions with mean zero has one degree of freedom.
- We study the case of $F$ having an "equidistant support"

$$
\{a-2 b, a-b, a\} .
$$

For $x>0$, define a "measure of non-integrity"

$$
\lceil x\rfloor=\min \left\{\frac{\lceil x\rceil}{x}-1,1-\frac{\lfloor x\rfloor}{x}\right\} \in[0,1] .
$$

Obviously $\lceil x\rfloor=0 \Leftrightarrow x \in \mathbb{N}$.

## Tri-atomic Distributions

## Theorem 4

Suppose that $F=\left(f_{1}, f_{2}, f_{3}\right)$ is a tri-atomic distribution with zero mean supported in $\{a-2 b, a-b, a\}, \epsilon>0$ and $a \leq b$. Then $F \in \mathcal{D}_{2}$ if and only if it is the following three cases.
(i) $a=b$ and $f_{2} \geq\left\lceil\frac{1}{b}\right\rfloor$.
(ii) $a<b$ and $\frac{1}{b} \in \mathbb{N}$.
(iii) $a<b, \frac{1}{b}-\frac{1}{2} \in \mathbb{N}$ and $f_{2} \geq \frac{a}{2}$.

- cf. Theorem 3 (condition $2 / b \in \mathbb{N}$ )


## Tri-atomic Distributions

The corresponding distributions in Theorem 4:
(i) $\left(f_{1}, f_{2}, f_{3}\right) \in \operatorname{cx}\left\{(0,1,0), \frac{1}{2}\left(1-\left\lceil\frac{1}{b}\right\rfloor, 2\left\lceil\frac{1}{b}\right\rfloor, 1-\left\lceil\frac{1}{b}\right\rfloor\right)\right\}$.
(ii) $\left(f_{1}, f_{2}, f_{3}\right) \in \operatorname{cx}\left\{\left(0, \frac{a}{b}, 1-\frac{a}{b}\right), \frac{1}{2}\left(\frac{a}{b}, 0,2-\frac{a}{b}\right)\right\}$.
(iii) $\left(f_{1}, f_{2}, f_{3}\right) \in \operatorname{cx}\left\{\left(0, \frac{a}{b}, 1-\frac{a}{b}\right), \frac{1}{2}\left(\frac{a}{b}-\frac{a}{2}, a, 2-\frac{a}{b}-\frac{a}{2}\right)\right\}$.

## Progress of the Talk

## (1) Question

(2) Some Examples
(3) Some Answers
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## Some More to Expect

- It is possible to further characterize $n$-atomic distributions with an equidistant support (things get ugly though).
- We guess: for any distribution $F$
- with an equidistant support, or
- with finite density and a bounded support, there exists a number $M>0$ such that $F \in \mathcal{D}_{2}(\mathrm{U}[-m, m], \mathrm{U}[-m, m])$ for all $m \in \mathbb{N}$ and $m>M$.


## Some More to Think

- Two uniforms with different lengths?
- Three or more uniform distributions?
- Other types of distributions?
- Applications?

We yet know very little about the problem of $\mathcal{D}_{2}$

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## Danke Schön



## Thank you for your kind attention

