# Probability Distributions: New Results on their Moment Determinacy 

Jordan Stoyanov (stoyanovj@gmail.com)

IMI BAS \& Newcastle University (UK)

National Colloquium in Mathematics
UBM \& IMI-BAS, Sofia, BULGARIA
29 November 2017, 16:15

## Brief Historical Comments:

Chebyshev: proved the C.L.T. for arbitrary seq. of random variables. He showed that the moments of the centered normalized sums convergence to those of the normal distribution $\mathcal{N}(0,1)$. Hence ... job done!

Markov: refined the proof, asking the Teacher if it is possible to have more than one distribution corresponding to a moment sequence.

Chebyshev sent this to Hermite, who ... passed to Stieltjes all Chebyshev stuff (papers, ideas, continued fractions, integral inequalities) Stieltjes did a great job! 1894 Name, "iff" conditions, two examples of distributions on $\mathbb{R}_{+}$which are non-unique in terms of the moments. Later, strong interest: Hamburger, Hardy, Carleman, Riesz, Fréchet, Cramér, Hausdorff, Akhiezer, Krein, Nudelman, Karlin, ...

Probability Theory ... or Measures and Functions
Probability space $(\Omega, \mathcal{F}, \mathbf{P})$, random events, variables, vectors, etc.
Random variable $X$, range of values $\mathbb{R}, \mathbb{R}_{+}$of a subset.
Distribution function $F(x)=\mathbf{P}[\omega: X(\omega) \leq x], x \in \mathbb{R}$. Properties, supp.
Density $f(x)=F^{\prime}(x), x \in \mathbb{R}$ (not always !)
Distribution $\mu=\mu_{F}$, the measure on $(\mathbb{R}, \mathcal{B})$ generated by $F$.
Characteristic function $\psi(t):=\mathbf{E}\left[\mathrm{e}^{\mathrm{i} t X}\right], t \in \mathbb{R}$ (Fourier Transform, FT)
Moment generating function If $M(t)=\mathbf{E}\left[\mathrm{e}^{t X}\right]<\infty, t \in\left(-t_{0}, t_{0}\right), t_{0}>0$
Bellow we use standard abbreviations: r.v., d.f., ch.f., m.g.f.

Kolmogorov Theorem: Given a d.f. $G$, then there exists a probability space and a r.v. $Y$ such that the d.f. of $Y$ is exactly $G$.

Given a r.v. $X \sim F$, expectation (expected value, mean value, $\ldots$ ) is:
$\mathbf{E}[X]=\int_{\Omega} X(\omega) \mathrm{d} \mathbf{P}(\omega)=\int x \mathrm{~d} F(x)=\int x f(x) \mathrm{d} x=\int x \mu(\mathrm{~d} x)$.
Assume $X$ has absolute moments: $\mathbf{E}\left[|X|^{k}\right]<\infty, k=1,2, \ldots$. Then $m_{k}=\mathbf{E}\left[X^{k}\right]$, moment of order $k ;\left\{m_{k}, k=1,2, \ldots\right\}$ moment seq. of $F, X$.

Question: We have $\left\{m_{k}\right\}$. Is there a d.f. $G \neq F$ with the same moments? Answer: Sometime 'yes', sometime 'no'.

Equivalently: If $f(x)$ is a function and $\hat{f}(t)$ its FT , can we recover $f$ if we know all derivatives of $\hat{f}(t)$ at $t=0$ ? Recall: $\left.\frac{d^{k}}{d t^{k}} \hat{f}(t)\right|_{t=0}=m_{k}$. Notice, $\left\{m_{k}\right\}$ is a discrete info, $f$ and $F$ are infinite ... Inverse Fourier. Numerical methods.

Terminology: Name of the moment problem, depends on the $\operatorname{supp}(F)$ :
$[0,1]$ (Hausdorff); $\mathbb{R}_{+}=[0, \infty)$ (Stieltjes); $\mathbb{R}=(-\infty, \infty)$ (Hamburger)
Fact 1: For any $X, F, f, \mu$ with finite moments $\left\{m_{k}\right\}$ there are two possibilities, (a) or (b):
(a) $F$ is $M$-determinate, or unique with these moments ( $M$-det), i.e.

$$
\int x^{k} \mathrm{~d} F(x)=\int x^{k} \mathrm{~d} G(x) \text { for all } k=1,2, \ldots \Rightarrow F=G
$$

(b) $F$ is $M$-indeterminate ( $M$-indet) if there is at least one d.f. $G$ with

$$
\int x^{k} \mathrm{~d} F(x)=\int x^{k} \mathrm{~d} G(x) \text { for all } k=1,2, \ldots, \text { but } G \neq F \text {. }
$$

Fact 2: For any M-indet $F$, there are infinitely many absolutely continuous distributions, infinitely many discrete distributions, and infinitely many singular distributions, all having the same moments as $F$.

The latter can be extracted from a series of works by C. Berg and co.
Notion: Stieltjes class. See next page.
Importance of $\mathbf{M}$-uniqueness: Fréchet-Shohat Theorem:
Sequence of measures $\mu_{n}, n=1,2, \ldots$ s.t. for each $k, \int x^{k} \mathrm{~d} \mu_{n} \rightarrow m_{k}$ as $n \rightarrow \infty$. Then $\left\{m_{k}\right\}$ is the moment sequence of a measure, say $\mu$, and if $\mu$ is M-det, then $\mu_{n}$ converges weakly to $\mu$.
Wide applications in graph theory, number theory, statistical physics, ..., combinatorics.

Statistical Inference: Recall skewness and kurtosis coefficients. Risky business. Optimization problems, including in Financial Mathematics.

Log-normal distribution:
$Z \sim \mathcal{N}(0,1), \varphi(x)=\frac{1}{\sqrt{2 \pi}} \mathrm{e}^{-x^{2} / 2}, x \in \mathbb{R}, m_{2 k+1}=0, m_{2 k}=(2 k-1)!!$.
Then $X=\mathrm{e}^{Z} \sim \log \mathcal{N}(0,1)$, density

$$
f(x)=\frac{1}{\sqrt{2 \pi}} \frac{1}{x} \exp \left[-\frac{1}{2}(\ln x)^{2}\right], x>0 ; f(x)=0, x \leq 0 .
$$

About $X$ : No m.g.f., HT, $m_{k}=\mathbf{E}\left[X^{k}\right]=e^{k^{2} / 2}, k=1,2, \ldots$
Two infinite sets of r.v.s, one absolutely continuous, one discrete:
$X_{\varepsilon}, \varepsilon \in[-1,1]:$ density $f_{\varepsilon}(x)=f(x)[1+\varepsilon \sin (2 \pi \ln x)], x>0$
$Y_{a}, a>0: \mathbf{P}\left[Y_{a}=a e^{n}\right]=a^{-n} \mathrm{e}^{-n^{2} / 2} / A, n=0, \pm 1, \pm 2, \ldots$
Shocking property: $\quad \mathbf{E}\left[X_{\varepsilon}^{k}\right]=\mathbf{E}\left[Y_{a}^{k}\right]=\mathbf{E}\left[X^{k}\right]=\mathrm{e}^{k^{2} / 2}, k=1,2, \ldots$.
Conclusion: LogN is M-indet! So 'many' others, the same moments.

Classical Conditions: Back to Stieltjes and Hamburger: moment seq. $\left\{m_{k}, k=1,2, \ldots\right\}$, write an infinite seq. of Hankel matrices: $H_{N}$ involves $m_{0}=1, m_{1}, \ldots, m_{N}, \ldots, m_{N}, \ldots, m_{2 N}, N=1,2 \ldots$
If all $H_{N}$ are non-negative, or positive-definite, then the is a measure $\mu$ with these moments, and it is unique. Advanced case (O.K., Ark. Math. 2010): complex-valued, pseudo-positive-definite ...

Recently, Berg, Chen, Ismail (Math. Scand. 2002): If $\lambda_{N}$ is the smallest eigenvalue of $H_{N}$, then $\lambda_{N} \searrow$ as $N \rightarrow \infty$. Moreover, there is a unique measure $\mu$ with the moments $\left\{m_{k}\right\}$ iff $\lambda_{N} \rightarrow 0$ as $N \rightarrow \infty$.

Not easy to check. For a century, find checkable conditions, even if they are only sufficient, or only necessary for either M-det or M-indet.

All is done!? MathSciNet, 2017="moment problem", 226 new papers.

Cramér condition: For a r.v. $X \sim F$ on $\mathbb{R}$, let the m.g.f. exists:

$$
M(t)=\mathbf{E}\left[\mathrm{e}^{t X}\right]<\infty, t \in\left(-t_{0}, t_{0}\right), t_{0}>0 \text { (light tails). }
$$

- $X$ has all moments finite; - $X$, i.e. $F$, is $M$-det.

Remark: Existence of m.g.f. $\Leftrightarrow$ analytic ch.f. More, strong Cramér: $M(t)<\infty$ for all $t \in \mathbb{R} \Leftrightarrow$ real analytic ch.f.

Examples: Bounded supp $(F)$ (Hausdorff case), Exp, $\mathcal{N}$, Laplace.
If $F$ has heavy tail(s) (no m.g.f.), $F$ is either $M$-det or $M$-indet
Question: When is a heavy tailed $F$ unique? And when non-unique?
We will see that we may have uniqueness even if the tail(s) are 'heavy', but not too much, and non-uniqueness comes out if the tail(s) of the distribution is 'very heavy'. Later, example with Exp.
"New" Condition: G. Hardy (1917/1918). The Math. Messenger Statement (Hardy): r.v. $X>0, X \sim F$. Suppose $\sqrt{X}$ has m.g.f.:
$\mathbf{E}\left[\mathrm{e}^{t \sqrt{x}}\right]<\infty$ for $t \in\left[0, t_{0}\right), t_{0}>0 \quad(H)=$ Hardy's condition; $\frac{1}{2}$-Cramér.
Then all $m_{k}=\mathbf{E}\left[X^{k}\right]<\infty, k=1,2, \ldots$ and $F$ is the only d.f. with $\left\{m_{k}\right\}$.
Hint: Details in S\&L TPA (2012/2013). The main, $F$ is arbitrary. More, condition $(\mathrm{H}) \Longleftrightarrow m_{k}(X) \leq c^{k}(2 k)!\Rightarrow \mathrm{C}\left[\left\{m_{k}\right\}\right]=\infty \Rightarrow X$ is M-det.

Notice: The condition is on $\sqrt{X}$ but the conclusion is for $X$.
Corollary: If a r.v. $X>0$ has a m.g.f., then its square $X^{2}$ is M -det.
Result: $\ln (H), \frac{1}{2}$ is the best possible constant for $X$ to be M-det. For each $\rho \in\left(0, \frac{1}{2}\right)$ there is a r.v. $Y$ with $\mathbf{E}\left[\mathrm{e}^{t Y^{\rho}}\right]<\infty$ s.t. $Y$ is M -indet.
Comment: Hardy's condition is sufficient but not necessary for M-det.

Carleman: Quasi-analytic functions, book 1926.
Known are all moments $m_{k}=\mathbf{E}\left[X^{k}\right], k=1,2, \ldots$
For $X$ in $\mathbb{R}$ or $\mathbb{R}_{+}$, define the following Carleman quantity:

$$
\mathrm{C}=\sum_{k=1}^{\infty} \frac{1}{\left(m_{2 k}\right)^{1 / 2 k}}, \quad \mathrm{C}=\sum_{k=1}^{\infty} \frac{1}{\left(m_{k}\right)^{1 / 2 k}} .
$$

Theorem 1 (Carleman): $\mathrm{C}=\infty \Rightarrow F$ is M-det. (Only sufficient!)
Remark: If $F$ is M -indet, then necessarily $\mathrm{C}<\infty$ (converse Carleman's).
Theorem 2: (Koosis and Pakes) Suppose $C<\infty$. If $F$ has a density $f$ such that $-\ln f\left(\mathrm{e}^{x}\right)$ is convex for $x>x_{0}>0$, then $F$ is M -indet.

Krein: Let $X \sim F$, density $f>0$, finite moments. For $X$ in $\mathbb{R}$ or $\mathbb{R}_{+}$, define the Krein quantity (Krein 1944 and Slud 1993):

$$
\mathrm{K}[f] \equiv \int_{-\infty}^{\infty} \frac{-\ln f(y)}{1+y^{2}} \mathrm{~d} y, \quad \mathrm{~K}[f] \equiv \int_{a}^{\infty} \frac{-\ln f\left(y^{2}\right)}{1+y^{2}} \mathrm{~d} y, a \geq 0 .
$$

Theorem 1: $\mathrm{K}[f]<\infty \Rightarrow F$ is M -indet. (Only sufficient!)
Remark: If $F$ is M -det, then necessarily $\mathrm{K}[f]=\infty$ (converse Krein's).
We say that Lin's condition is satisfied if $f>0, f$ is smooth and

$$
L_{f}(x):=\frac{-x f^{\prime}(x)}{f(x)} \not \nearrow \infty \text { as } x_{0}<x \rightarrow \infty .
$$

Theorem 2: (Lin 1997) $\mathrm{K}[f]=\infty+$ Lin's condition $\Rightarrow F$ is M-det.
Remark: There are variations and extensions (Pedersen, Berg, ...).

## Rate of Growth of the Moments and (In)Determinacy:

Stieltjes case: $X \sim F$, unbounded $\operatorname{supp}(F) \subset \mathbb{R}_{+}, \quad m_{k}=\mathbf{E}\left[X^{k}\right]$.
Property 1: (i) $\left\{m_{k}\right\}$ is log-convex. (ii) As $k \rightarrow \infty, m_{k} \not \subset \infty$. How much lager $m_{k+1}$ is compared with $m_{k}$ ? Involve the ratio

$$
\Delta_{k}=\frac{m_{k+1}}{m_{k}}, k=1,2, \ldots \Rightarrow \Delta_{k} \nearrow \text { strictly } \Rightarrow \text { unique } \lim _{k \rightarrow \infty} \Delta_{k} .
$$

Property 2: Case $\lim _{k \rightarrow \infty} \Delta_{k}=c_{*}=$ const $<\infty \Leftrightarrow X$ is bounded.
Assume now $\Delta_{k} \nearrow \infty$ is $R V(\delta)$, some $\delta \geq 0$ (in Karamata's sense):

$$
\Delta_{k} \approx k^{\delta} \ell_{k} \text { for large } k \text {, with a monotone }\left\{\ell_{k}\right\} \in S V \text {. }
$$

Def.: $\delta=$ rate of growth of the moments of $X$. We have $0 \leq \delta \leq \infty$.
Hamburger case: Now $X \sim F, \operatorname{supp}(F) \subset \mathbb{R}$, we work with even order moments. As above we have $m_{2 k} \nearrow \infty$, etc., involve

$$
\Delta_{k}=\frac{m_{2 k+2}}{m_{2 k}} \approx k^{\delta} \ell_{k} \quad \text { as } k \rightarrow \infty
$$

Remark: Two boundary cases. If $X \in[0,1]$ or $[-1,1]$, moments $\searrow$, easy to see $\delta=0$. Take $X \sim \log \mathcal{N}, m_{k}=\mathrm{e}^{k^{2} / 2}, \Delta_{k}=m_{k+1} / m_{k}=\mathrm{e}^{k+1 / 2} \nearrow+\infty$ as $k \rightarrow \infty$ faster than any power of $k$, hence $\delta=\infty$.

Three statements for both cases (Stieltjes and Hamburger):
Statement 1: If $\delta \leq 2$, then $X$ is M -det.
Statement 2: $\delta=2$ is the best possible constant for which $X$ is M-det. Equiv: There is a r.v. $Y$ with $\Delta_{k} \approx k^{2+\varepsilon} \ell_{k}, \varepsilon>0$, s.t. $Y$ is M -indet.

Statement 3: If $\delta>2$ and Lin's condition holds, then $X$ is M -indet.
Remark: Interestingly, Lin's condition is essentially involved with the converse Krein to get M-det, and here, with 'fast' rate $\delta$, to get M-indet.

Example 1: $\xi \sim \operatorname{Exp}(1), \mathrm{e}^{-x}, x>0, m_{k}=k!$, m.g.f. $=1 /(1-t), t<1$.
Clearly, $\xi$ is M-det. And we know everything for any power $\xi^{r}, r>0$.
$\mathbf{r}=2: \quad \xi^{2}, \quad m_{k}=(2 k)!$, fast $\pi$, still $\xi^{2}$ is M-det. Notice, no m.g.f.!
How to show that $f(x)=\frac{1}{2} x^{-1 / 2} \exp \left(-x^{-1 / 2}\right), x>0$ is the only solution of

$$
\int_{0}^{\infty} x^{k} f(x) \mathrm{d} x=(2 k)!, k=1,2, \ldots ?
$$

$\mathbf{r}=3: \xi^{3}, m_{k}=(3 k)!$, very fast $\pi, \xi^{3}$ is M -indet. How to show that

$$
\int_{0}^{\infty} x^{k} f(x) \mathrm{d} x=(3 k)!, k=1,2, \ldots
$$

has infinitely many solutions? Take, e.g. the Stieltjes class

$$
f_{\varepsilon}(x)=f(x)[1+\varepsilon h(x)], x>0, \varepsilon \in[-1,1]
$$

where $f(x)=\frac{1}{3} x^{-2 / 3} \mathrm{e}^{-x^{1 / 3}}$ and $h(x)=\sin \left(\frac{\pi}{6}-\sqrt{3} x^{1 / 3}\right)$.

Example 2: $Z \sim \mathcal{N}(0,1), \frac{1}{\sqrt{2 \pi}} \mathrm{e}^{-x^{2} / 2}, m_{2 k+1}=0, m_{2 k}=(2 k-1)!!$.
We have, e.g., $Z$ is M -det, $Z^{2}$ is M -det, $Z^{4}$ is M -det, but $Z^{3}$ is M -indet.
Strange Case: While $X=Z^{3}$ is M -indet, it turns out, $|X|$ is M -det.
Hint: $X$ is in $\mathbb{R}$ (Hamburger), $|X|$ in $\mathbb{R}_{+}$(Stieltjes). Density of $X$ :
$g_{X}(x)=\frac{1}{3 \sqrt{\pi}}|x|^{-2 / 3} \exp \left[-|x|^{2 / 3}\right], x \in \mathbb{R} \Rightarrow \mathrm{~K}\left[g_{x}\right]<\infty, X$ is M -indet.
Or: $m_{2 k}(X)=(6 k-1)!$ !, rate $\delta_{X}=3$, Lin's cond. holds $\Rightarrow X$ is M-indet.
Stieltjes class: For some $h, \mathbf{S}(g, h)=\left\{g_{\varepsilon}=g[1+\varepsilon h], \varepsilon \in[-1,1]\right\}$.
Next, for $|Z|, \quad m_{k}=\mathbf{E}\left[|Z|^{k}\right]=\frac{2^{k / 2}}{\sqrt{\pi}} \Gamma\left(\frac{k+1}{2}\right), g_{|X|}(x)=2 g_{x}(x), x>0$, $\mathrm{K}\left[g_{|X|}\right]=\infty, g_{|X|}$ satisfies Lin's condition, hence $|X|$ is M -det.
Also: $m_{k}(|X|)=m_{3 k}(|Z|)=\frac{2^{3 k / 2}}{\sqrt{\pi}} \Gamma\left(\frac{3 k+1}{2}\right)$, rate $\delta_{|X|}=3 / 2<2,|X|$ is M -det.

Current topics (a few random choices from the boxes):

- Multivariate distributions/measures and related moment problems.
- How to write Lin's condition for discrete distributions?
- Class of distributions $F_{a}$, density $f_{a}$ satisfies 2 nd order ODE:

$$
f^{\prime \prime}(x)-a(x) f(x)=0 \quad(+ \text { initial conditions })
$$

Find $a(\cdot)$ so that $f$ is density, either Krein's condition, or converse Krein's

+ Lin's condition. Hence, a r.v. $X \sim f$ is either M-indet or M-det.
All above are parts of a joint project with P. Kopanov.
- Distributions whose densities satisfy Kolmogorov PDEs, connection with Itô SDEs. Joint work with M. Zhitlukhin.
- More topics on functions, MaxEnt approach, Lévy processes, etc. with G.D. Lin, A. Tagliani, M. Milev, M. Savov, K. Penson.


## Few Final Comments. Open Questions (to me):

Question: Two measures, $\mu$ and $\nu$, both with all moments finite. Let

$$
\int x^{k} \mathrm{~d} \mu \leq \int x^{k} \mathrm{~d} \nu \text { for all } k=1,2, \ldots \text { (M-domination). }
$$

(1) We know that $\mu$ is M -indet, does this imply that $\nu$ is also M -indet?
(2) We know that $\nu$ is M -det, does this imply that $\mu$ is also M -det?

Open Question: In dim. $n \geq 2$, how to write Krein's condition?
Open Question: Find discrete r.v. on $\mathbb{R}_{+}$with moments $\{(3 k)$ !\}. Equivalently: Find reals, $x_{1}<x_{2}<\ldots<x_{n}<\ldots$ (unbounded) and $p_{1}, p_{2}, \ldots, p_{n}, \ldots$, all $>0$ with $\sum_{n=1}^{\infty} p_{n}=1$ (disc. probab. distr.) such that

$$
\sum_{n=1}^{\infty} x_{n}^{k} p_{n}=(3 k)!\text { for any } k=1,2, \ldots .
$$

Open Question: Find discrete r.v. with $m_{2 k-1}=0, m_{2 k}=(6 k-1)!!$.

References: Books: Akhiezer/Krein 1938, Shohat/Tamarkin 1943, Akhiezer 1961/1965, Berg \& co. 1981, Vorobyev 1984, Lasserre 2010, and ... Schmüdgen, The Moment Problem, 13 Dec 2017

Hardy, GH (1917/1918): The Mathematical Messenger 46/47
Berg, C (1988): Ann. Probab. 16 910-913; JOTP (2005)
Pedersen, HL (1988): J. Approx. Theory 95 90-100
Lin, GD (1997): Statist. Probab. Letters 35 85-90
Pakes, A (2007): J. Math. Anal. Appl. 326 1268-1290.
Putinar, M, Schmüdgen, K (2008): Indiana Univ. Math. J. 57 2931-68
Kleiber, C, Stoyanov, J (2011/2013): J. Multivar. Analysis 113 7-18
Stoyanov, J, Lin, GD (2012): Theory Probab. Appl. 57 811-820
[SIAM edition: TPA 57 (2013) 699-708]
Lin, GD, Stoyanov, J (2014 - 2016): JOTP, JSPI, PMS
Stoyanov, J (2013): Counterexamples in Probability. 3rd edn.
Dover, New York. (1st ed, 2nd ed: Wiley 1987, 1997)

