



Eigenvalue is not a dirty word

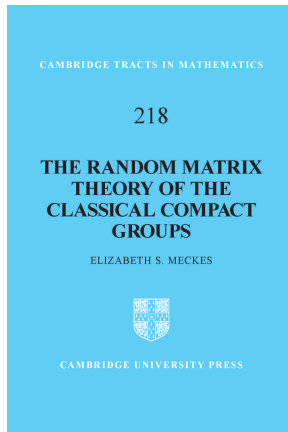
My mathematical collaborations with Elizabeth Meckes

Mark Meckes

Case Western Reserve University
Currently visiting Oxford University

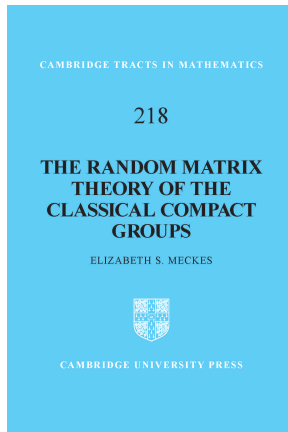
Southeastern Probability Conference
17 May 2021

From the preface of Elizabeth's book:



If my writing helps to illuminate the ideas I have tried to describe, it is because I got to talk it out first at the breakfast table.

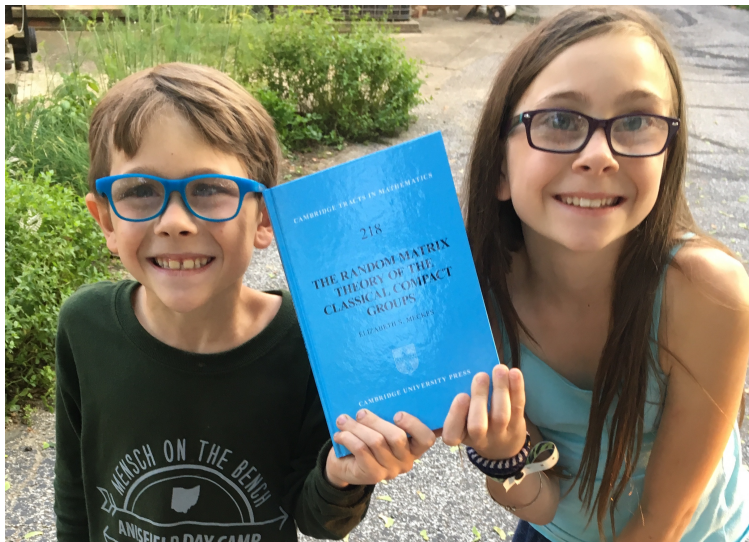
From the preface of Elizabeth's book:



If my writing helps to illuminate the ideas I have tried to describe, it is because I got to talk it out first at the breakfast table.



We weren't the only ones at that breakfast table:



Peter and Juliette

You may have met those two before

You may have met those two before



at a conference,

You may have met those two before



at a conference,



or an
AMS
meeting,

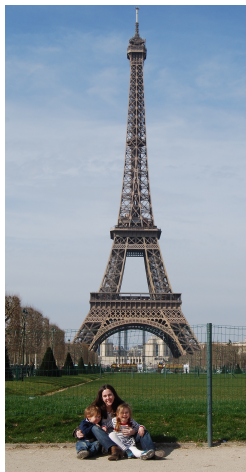
You may have met those two before



at a conference,



or an
AMS
meeting,



or when
we went
some-
where to
give
seminar
talks.

Meckes family fun fact

Juliette got her first passport as a baby to go with us to a conference in Banff...



Advances in Stochastic Inequalities and Their
Applications

Meckes family fun fact

Juliette got her first passport as a baby to go with us to a conference in Banff...

and so did Peter.



Advances in Stochastic Inequalities and Their Applications



High Dimensional Probability VI

Meckes family fun fact

Juliette got her first passport as a baby to go with us to a conference in Banff...

and so did Peter.



Advances in Stochastic Inequalities and Their Applications



High Dimensional Probability VI

(This wasn't the first conference for either of them.)

Elizabeth grew up around academic shop talk:

ISSN 0036-8075

27 July 1984

Volume 225, No. 4660

SCIENCE

Intestinal Uptake and Metabolism of Auranofin, A New Oral Gold-Based Antiarthritis Drug

KATHERINE TEPPERMAN*

RICHARD FINER

STUART DONOVAN

*Department of Biological Sciences,
University of Cincinnati,
Cincinnati, Ohio 45221*

R. C. ELDER, J. DOI

DAVID RATLIFF, KIN NG

*Department of Chemistry,
University of Cincinnati*

Elizabeth grew up around academic shop talk:

ISSN 0036-8075
27 July 1984
Volume 225, No. 4660

SCIENCE

Intestinal Uptake and Metabolism of Auranofin, A New Oral Gold-Based Antiarthritis Drug

KATHERINE TEPPERMAN*

RICHARD FINER

STUART DONOVAN

*Department of Biological Sciences,
University of Cincinnati,
Cincinnati, Ohio 45221*

R. C. ELDER, J. DOI

DAVID RATLIFF, KIN NG

*Department of Chemistry,
University of Cincinnati*

Katherine Tepperman = "Mom"

Elizabeth grew up around academic shop talk:

ISSN 0036-8075
27 July 1984
Volume 225, No. 4660

SCIENCE

Intestinal Uptake and Metabolism of Auranofin, A New Oral Gold-Based Antiarthritis Drug

KATHERINE TEPPERMAN*

RICHARD FINER

STUART DONOVAN

*Department of Biological Sciences,
University of Cincinnati,
Cincinnati, Ohio 45221*

R. C. ELDER, J. DOI

DAVID RATLIFF, KIN NG

*Department of Chemistry,
University of Cincinnati*

Katherine Tepperman = “Mom”

R.C. Elder = “Dad”

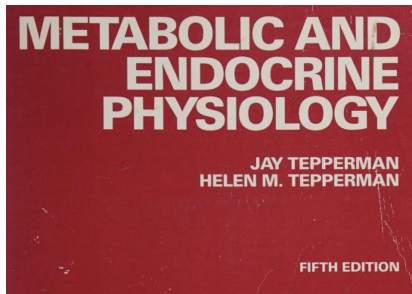
It goes back farther than that:

**FOURIER EXPANSIONS OF DOUBLY PERIODIC
FUNCTIONS OF THE THIRD KIND.***

BY JOHN D. ELDER.

Annals of Mathematics, vol. 31 no. 4, Oct. 1930, pp. 641–654

John D. Elder = “Grandpa John”



Jay Tepperman =
“Grandpa Jay”

Helen M. Tepperman =
“Grandma”

The dark side to this kind of background...

Young Katherine Tepperman automatically stopped listening to the dinner table conversation whenever she heard “glucose-6-phosphate”.

Role of hormones in glucose-6-phosphate dehydrogenase adaptation of rat liver¹

HELEN M. TEPPERMAN AND JAY TEPPERMAN (With the Technical Assistance of Janet M. DeWitt and Andrew Branch)
Department of Pharmacology, State University of New York, Upstate Medical Center, Syracuse, New York

American Journal of Physiology, vol. 202 no. 3, March 1962, pp. 401–406

The dark side to this kind of background...

Young Katherine Tepperman automatically stopped listening to the dinner table conversation whenever she heard “glucose-6-phosphate”.

Role of hormones in glucose-6-phosphate dehydrogenase adaptation of rat liver¹

HELEN M. TEPPERMAN AND JAY TEPPERMAN (With the Technical Assistance of Janet M. DeWitt and Andrew Branch)
Department of Pharmacology, State University of New York, Upstate Medical Center, Syracuse, New York

American Journal of Physiology, vol. 202 no. 3, March 1962, pp. 401–406

She had to unlearn this habit on her way to becoming a biologist!

The dark side to this kind of background...

Young Elizabeth Elder (later Meckes) tuned out when she heard “Na, K-ATPase”.

Biophysical Journal Volume 79 August 2000 853–862

Combined Allosteric and Competitive Interaction between Extracellular Na⁺ and K⁺ During Ion Transport by the α_1 , α_2 , and α_3 Isoforms of the Na, K-ATPase

David M. Balshaw,* Lauren A. Millette,[†] Katherine Tepperman,[†] and Earl T. Wallick*

The dark side to this kind of background...

Young Elizabeth Elder (later Meckes) tuned out when she heard “Na, K-ATPase”.

Biophysical Journal Volume 79 August 2000 853–862

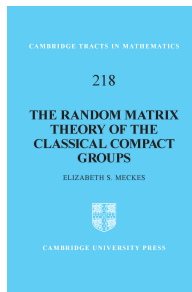
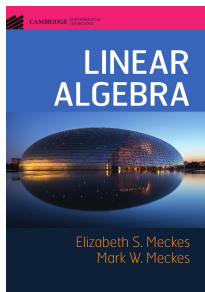
Combined Allosteric and Competitive Interaction between Extracellular Na⁺ and K⁺ During Ion Transport by the α_1 , α_2 , and α_3 Isoforms of the Na, K-ATPase

David M. Balshaw,* Lauren A. Millette,[†] Katherine Tepperman,[†] and Earl T. Wallick*

She sidestepped having this as a potential problem by going into pure mathematics.

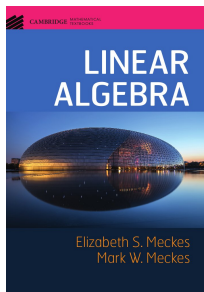
The next generation

Juliette says she tunes out when she hears “eigenvalue”.

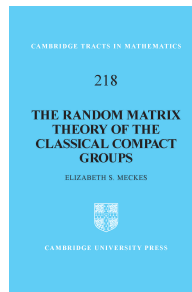


The next generation

Juliette says she tunes out when she hears “eigenvalue”.



eigenvalue, 69, 69–73, 75, 122
and determinant, 362
and trace, 362
geometric multiplicity, 214, 365
multiplicity, 361, 360–362, 364
of a diagonal matrix, 71, 122
of a self-adjoint map linear map, 314
of a unitary matrix, 286
of an upper triangular matrix, 215, 220, 361
of similar matrices, 207
of transpose matrix, 179
root of characteristic polynomial, 358



3	Eigenvalue Distributions: Exact Formulas	60
3.1	The Weyl Integration Formula	60
3.2	Determinantal Point Processes	71
3.3	Matrix Moments	80
3.4	Patterns in Eigenvalues: Powers of Random Matrices	85
4	Eigenvalue Distributions: Asymptotics	90
4.1	The Eigenvalue Counting Function	90
4.2	The Empirical Spectral Measure and Linear Eigenvalue Statistics	106
4.3	Patterns in Eigenvalues: Self-Similarity	113
4.4	Large Deviations for the Empirical Spectral Measure	117

It begins

2004: Elizabeth was a PhD student at Stanford and I was a lecturer.

It begins

2004: Elizabeth was a PhD student at Stanford and I was a lecturer.

Persi Diaconis gave a series of lectures on [Stein's method](#).

It begins

2004: Elizabeth was a PhD student at Stanford and I was a lecturer.

Persi Diaconis gave a series of lectures on [Stein's method](#).

These led Elizabeth directly into her thesis work and collaboration with Sourav Chatterjee.

It begins

2004: Elizabeth was a PhD student at Stanford and I was a lecturer.

Persi Diaconis gave a series of lectures on [Stein's method](#).

These led Elizabeth directly into her thesis work and collaboration with Sourav Chatterjee.

I also went to those lectures and had an idea how to apply Stein's method to the [central limit problem for convex bodies](#):

Are marginals of the uniform measure on a high-dimensional convex body typically almost Gaussian?

It begins

2004: Elizabeth was a PhD student at Stanford and I was a lecturer.

Persi Diaconis gave a series of lectures on [Stein's method](#).

These led Elizabeth directly into her thesis work and collaboration with Sourav Chatterjee.

I also went to those lectures and had an idea how to apply Stein's method to the [central limit problem for convex bodies](#):

Are marginals of the uniform measure on a high-dimensional convex body typically almost Gaussian?

and quickly got stuck.

It begins

I asked Elizabeth for help and the result was:

“Theorem” (E. and M. Meckes, 2007)

If n is big and a probability measure μ on \mathbb{R}^n has certain kinds of symmetries, and satisfies some other hypotheses, then certain (one-dimensional) marginals of μ are quantifiably almost Gaussian.

It begins

I asked Elizabeth for help and the result was:

“Theorem” (E. and M. Meckes, 2007)

If n is big and a probability measure μ on \mathbb{R}^n has certain kinds of symmetries, and satisfies some other hypotheses, then certain (one-dimensional) marginals of μ are quantifiably almost Gaussian.

Remarks:

- Think of this as a Berry–Esseen-type theorem with stochastic independence replaced by a geometric symmetry condition.

It begins

I asked Elizabeth for help and the result was:

“Theorem” (E. and M. Meckes, 2007)

If n is big and a probability measure μ on \mathbb{R}^n has certain kinds of symmetries, and satisfies some other hypotheses, then certain (one-dimensional) marginals of μ are quantifiably almost Gaussian.

Remarks:

- Think of this as a Berry–Esseen-type theorem with stochastic independence replaced by a geometric symmetry condition.
- The “other hypotheses” are now known to hold for uniform measures on convex bodies.

It begins

I asked Elizabeth for help and the result was:

“Theorem” (E. and M. Meckes, 2007)

If n is big and a probability measure μ on \mathbb{R}^n has certain kinds of symmetries, and satisfies some other hypotheses, then certain (one-dimensional) marginals of μ are quantifiably almost Gaussian.

Remarks:

- Think of this as a Berry–Esseen-type theorem with stochastic independence replaced by a geometric symmetry condition.
- The “other hypotheses” are now known to hold for uniform measures on convex bodies.
- Elizabeth and I each wrote follow-ups to this work but didn’t work together on this again.

Some time (and one child) later...

“Theorem” (Chatterjee–Ledoux, 2009)

*If $1 \ll k \leq n$ and A is an $n \times n$ Hermitian matrix, then the **eigenvalues** of almost all $k \times k$ principal submatrices of A are almost the same.*

Some time (and one child) later...

“Theorem” (Chatterjee–Ledoux, 2009)

*If $1 \ll k \leq n$ and A is an $n \times n$ Hermitian matrix, then the **eigenvalues** of almost all $k \times k$ principal submatrices of A are almost the same.*

Elizabeth and I realized we could prove a coordinate-free version of this by combining some basic matrix analysis with tools she was using to study random marginals:

“Theorem” (E. and M. Meckes, 2011)

*If $1 \ll k \leq n$ and A is an $n \times n$ Hermitian matrix, then the **eigenvalues** of almost all k -dimensional **compressions** of A are almost the same.*

A little while after that...

Elizabeth improved her methods and thus her results for random marginals ([measure-theoretic Dvoretzky theorem](#)).

A little while after that...

Elizabeth improved her methods and thus her results for random marginals ([measure-theoretic Dvoretzky theorem](#)).

We applied the same improvements to **eigenvalues** of random compressions, then extended the same methods to **eigenvalues** of other random matrix models.

A little while after that...

Elizabeth improved her methods and thus her results for random marginals ([measure-theoretic Dvoretzky theorem](#)).

We applied the same improvements to **eigenvalues** of random compressions, then extended the same methods to **eigenvalues** of other random matrix models.

For example:

Theorem (E. and M. Meckes, 2013)

Let U be an $n \times n$ random matrix uniformly chosen from any of the classical compact matrix groups. Then

$$\mathbb{P} \left[W_1(\mu_U, \nu) \geq Cn^{-2/3} + t \right] \leq e^{-cn^2 t^2},$$

*where $t > 0$, μ_U is the **spectral measure** of U , ν is the uniform measure on the circle, and W_1 denotes Wasserstein/Kantorovich distance.*

Summer 2011

The work for that paper was done during the summer of 2011.

Peter was born on July 20.

Summer 2011

The work for that paper was done during the [summer of 2011](#).

Peter was born on [July 20](#).

Elizabeth wrote on her web page:

I think of this paper as being unofficially dedicated to our children: Peter, who stubbornly refused to be born while most of the work in this paper was done; and Juliette, who told me one morning that it would make her happy if I proved a theorem that day (I'm pretty sure it was what became Theorem 3.5).

Then the next year...

Sandrine Dallaporta used **determinantal point process** techniques to improve some of our results for **eigenvalues** of Gaussian random matrices.

Then the next year...

Sandrine Dallaporta used **determinantal point process** techniques to improve some of our results for **eigenvalues** of Gaussian random matrices.

We adapted her techniques to the classical compact groups,

Then the next year...

Sandrine Dallaporta used **determinantal point process** techniques to improve some of our results for **eigenvalues** of Gaussian random matrices.

We adapted her techniques to the classical compact groups, added in some cool results of Eric Rains,

Then the next year...

Sandrine Dallaporta used **determinantal point process** techniques to improve some of our results for **eigenvalues** of Gaussian random matrices.

We adapted her techniques to the classical compact groups, added in some cool results of Eric Rains, and mixed with log-Sobolev inequalities to prove:

Theorem (E. and M. Meckes, 2013)

Let U be an $n \times n$ random matrix uniformly chosen from any of the classical compact matrix groups and let $1 \leq m \leq n$. Then

$$\mathbb{P} \left[W_1(\mu_{U^m}, \nu) \geq C \frac{\sqrt{m(\log(n/m) + 1)}}{n} + t \right] \leq e^{-cn^2 t^2 / m}.$$

An observation about lemmas

A useful lemma often gets more citations than a big theorem.

An observation about lemmas

A useful lemma often gets more citations than a big theorem.

My most often-cited result in the last few years is probably a lemma from that last paper (concentration for Lipschitz functions of independent unitary random matrices)

An observation about lemmas

A useful lemma often gets more citations than a big theorem.

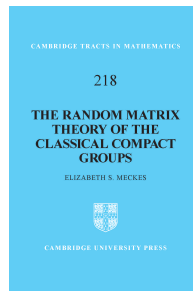
My most often-cited result in the last few years is probably a lemma from that last paper (concentration for Lipschitz functions of independent unitary random matrices), proved while we drove from Juliette's preschool to the university one morning.

An observation about lemmas

A useful **lemma** often gets more citations than a big **theorem**.

My most often-cited result in the last few years is probably a lemma from that last paper (**concentration for Lipschitz functions of independent unitary random matrices**), proved while we drove from Juliette's preschool to the university one morning.

But most of the recent citations go here:



Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

During sabbatical we worked together on:

- Comparison of metrics between log-concave distributions (motivated by the [central limit theorem for convex bodies](#))

Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

During sabbatical we worked together on:

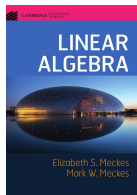
- Comparison of metrics between log-concave distributions (motivated by the [central limit theorem for convex bodies](#))
- Adapting the DPP techniques to quantify how uniformly distributed the [eigenvalues](#) of Ginibre random matrices are

Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

During sabbatical we worked together on:

- Comparison of metrics between log-concave distributions (motivated by the [central limit theorem for convex bodies](#))
- Adapting the DPP techniques to quantify how uniformly distributed the **eigenvalues** of Ginibre random matrices are

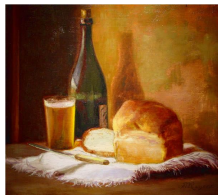


Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

During sabbatical we worked together on:

- Comparison of metrics between log-concave distributions (motivated by the [central limit theorem for convex bodies](#))
- Adapting the DPP techniques to quantify how uniformly distributed the [eigenvalues](#) of Ginibre random matrices are

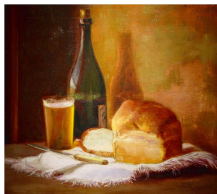


Sabbatical in Toulouse, 2013–14

By then it got to be habit to discuss whatever we were working on with each other, and we started writing the majority of our papers together.

During sabbatical we worked together on:

- Comparison of metrics between log-concave distributions (motivated by the [central limit theorem for convex bodies](#))
- Adapting the DPP techniques to quantify how uniformly distributed the [eigenvalues](#) of Ginibre random matrices are



1.1 Linear Systems of Equations

Bread, Beer, and Barley

We begin with a very simple example. Suppose you have 20 pounds of raw barley and you plan to turn some of it into bread and some of it into beer. It takes

Wrapping up

Our series of papers on convergence rates for **spectral measures** of random matrices finished up with:

Theorem (E. and M. Meckes, 2017)

Let U be an $n \times n$ random matrix uniformly chosen from the unitary group. Then

$$c \frac{\log n}{n} \leq \mathbb{E} d_K(\mu_U, \nu) \leq C \frac{\log n}{n},$$

where d_K denotes Kolmogorov distance.

again proved using DPP techniques.

Wrapping up

Our series of papers on convergence rates for **spectral measures** of random matrices finished up with:

Theorem (E. and M. Meckes, 2017)

Let U be an $n \times n$ random matrix uniformly chosen from the unitary group. Then

$$c \frac{\log n}{n} \leq \mathbb{E} d_K(\mu_U, \nu) \leq C \frac{\log n}{n},$$

where d_K denotes Kolmogorov distance.

again proved using DPP techniques.

So

$$d_K(\mu_U, \nu) \approx \frac{\log n}{n} \quad \text{but} \quad W_1(\mu_U, \nu) \lesssim \frac{\sqrt{\log n}}{n}.$$

Wrapping around

In the meantime, Elizabeth got interested in a conjecture of Coram and Diaconis:

*Take n consecutive **eigenvalues** of a random $2n \times 2n$ unitary matrix and stretch them around the whole unit circle. The result is statistically indistinguishable from the **eigenvalues** of a random $n \times n$ unitary matrix.*

Wrapping around

In the meantime, Elizabeth got interested in a conjecture of Coram and Diaconis:

*Take n consecutive **eigenvalues** of a random $2n \times 2n$ unitary matrix and stretch them around the whole unit circle. The result is statistically indistinguishable from the **eigenvalues** of a random $n \times n$ unitary matrix.*

Inspired by this we proved:

Theorem (E. and M. Meckes, 2016)

Let $N_{n,\theta}$ be the number of **eigenvalues** of an $n \times n$ random unitary matrix in an arc of length $\theta \in (0, \pi)$. Then

$$W_1(N_{n,\theta}, N_{2n,\theta/2}) \leq C\sqrt{n}\theta^2.$$

And coming full circle

There is a natural model of $n \times n$ Hermitian random matrices with prescribed **eigenvalues**.

And coming full circle

There is a natural model of $n \times n$ Hermitian random matrices with prescribed **eigenvalues**.

“Theorem” (E. and M. Meckes, 2020)

*Let Λ be a fixed $n \times n$ Hermitian matrix and let U be an $n \times n$ random matrix uniformly chosen from the unitary or orthogonal group. Then the marginals of the **entries** of the random matrix $U\Lambda U^*$ are quantifiably almost Gaussian.*

And coming full circle

There is a natural model of $n \times n$ Hermitian random matrices with prescribed **eigenvalues**.

“Theorem” (E. and M. Meckes, 2020)

*Let Λ be a fixed $n \times n$ Hermitian matrix and let U be an $n \times n$ random matrix uniformly chosen from the unitary or orthogonal group. Then the marginals of the **entries** of the random matrix $U\Lambda U^*$ are quantifiably almost Gaussian.*

We discussed perspectives and applications related to quantum information theory, free probability, and randomized linear algebra.

And coming full circle

This is a **quadratic** version of results from Elizabeth's thesis:

Theorem (E. Meckes, 2008)

Let U be an $n \times n$ random matrix uniformly chosen from the orthogonal group, and let A be a fixed $n \times n$ matrix with $\|A\|_{HS} = \sqrt{n}$. Then

$$d_{TV}(\text{Tr}(AU), N(0, 1)) \leq \frac{2\sqrt{3}}{n-1}.$$

(Multivariate versions appear in Chatterjee–E. Meckes, 2008.)

And coming full circle

This is a **quadratic** version of results from Elizabeth's thesis:

Theorem (E. Meckes, 2008)

Let U be an $n \times n$ random matrix uniformly chosen from the orthogonal group, and let A be a fixed $n \times n$ matrix with $\|A\|_{HS} = \sqrt{n}$. Then

$$d_{TV}(\text{Tr}(AU), N(0, 1)) \leq \frac{2\sqrt{3}}{n-1}.$$

(Multivariate versions appear in Chatterjee–E. Meckes, 2008.)

The newer theorem considers the distributions of random variables $\text{Tr}(AU \wedge U^*)$.

The proofs of both are via Stein's method.

Our penultimate (?) paper

Around 2007 we tossed around an idea for a proof using Stein's method, but it took us twelve years to figure out which theorem we should be using it to prove:

Our penultimate (?) paper

Around 2007 we tossed around an idea for a proof using Stein's method, but it took us twelve years to figure out which theorem we should be using it to prove:

“Theorem” (E. and M. Meckes, 2021)

*Let A be an $n \times n$ random matrix whose distribution is spherically symmetric. If p is a polynomial, then the distribution of the linear **eigenvalue** statistic $\text{Tr } p(A)$ is quantifiably almost Gaussian.*

To be continued...

We started working on another paper involving Stein's method and **eigenvalues** of random matrices in February 2020.

To be continued...

We started working on another paper involving Stein's method and **eigenvalues** of random matrices in February 2020.

An early observation from this project: most of the applications of the *Infinitesimal Version of Stein's Method of Exchangeable Pairs* are subsumed in a single normal approximation theorem which makes no mention of exchangeable pairs:

To be continued...

We started working on another paper involving Stein's method and **eigenvalues** of random matrices in February 2020.

An early observation from this project: most of the applications of the *Infinitesimal Version of Stein's Method of Exchangeable Pairs* are subsumed in a single normal approximation theorem which makes no mention of exchangeable pairs:

Proposition (Univariate version, for simplicity)

Suppose X is a uniform random point in a compact Riemannian manifold Ω , $f : \Omega \rightarrow \mathbb{R}$ is smooth, and $\lambda > 0$. Then

$$d_{TV}(f(X), N(0, 1)) \leq \frac{1}{\lambda} \mathbb{E} \left(|\Delta f(X) + \lambda f(X)| + \left| \|\nabla f(X)\|^2 - \lambda \right| \right).$$

To be continued...

This is a slight generalization of main results from two 2009 papers by Elizabeth.

To be continued...

This is a slight generalization of main results from two 2009 papers by Elizabeth.

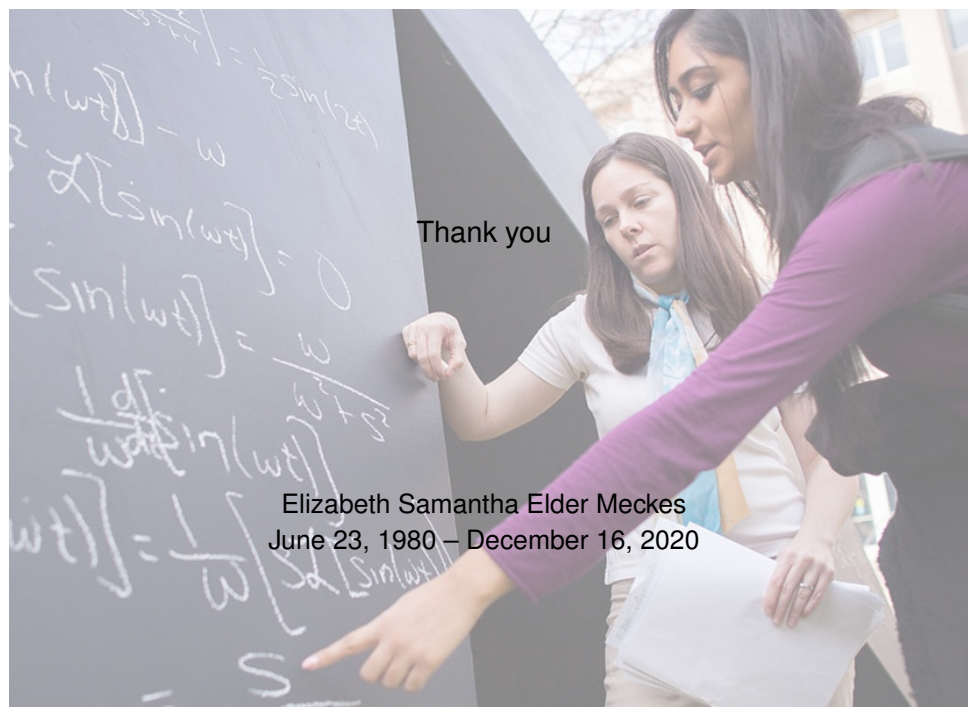
It is essentially proved (using exchangeable pairs) in those papers as well as in papers of Fulman (2009 and 2012) and Döbler and Stolz (2011).

To be continued...

This is a slight generalization of main results from two 2009 papers by Elizabeth.

It is essentially proved (using exchangeable pairs) in those papers as well as in papers of Fulman (2009 and 2012) and Döbler and Stolz (2011).

The main results of these papers (plus E. Meckes 2008 and E. and M. Meckes 2020) are all consequences of this master normal approximation lemma and its multivariate version.

A photograph of two women standing in front of a chalkboard. The woman on the left, wearing a white shirt and a blue patterned scarf, is pointing at the board. The woman on the right, wearing a purple long-sleeved shirt, is looking at the board. The chalkboard contains several mathematical equations written in white chalk. The text "Thank you" is overlaid on the right side of the image.

Thank you

Elizabeth Samantha Elder Meckes
June 23, 1980 – December 16, 2020