

# A calculus for the optimal quantification of uncertainties

Houman Owhadi

- Bayesian Brittleness.  
H. Owhadi, C. Scovel, T. Sullivan. 2013. arXiv:1304.6772
- Brittleness of Bayesian inference and new Selberg formulas.  
H. Owhadi, C. Scovel. 2013 arXiv:1304.7046
- Optimal Uncertainty Quantification. H. Owhadi, C. Scovel, T. Sullivan, M. McKerns and M. Ortiz. SIAM Review, 2013

## Kavli Royal Society 2014



# Reduction calculus

$$\mathcal{A} = \left\{ (f, \mu) \left| \begin{array}{l} f: \mathcal{X}_1 \times \cdots \times \mathcal{X}_m \rightarrow \mathbb{R}, \\ \mu = \mu_1 \otimes \cdots \otimes \mu_m, \\ \mathcal{G}(f, \mu) \leq 0 \end{array} \right. \right\}$$

$$\mathcal{G}(f, \mu) \leq 0 \Leftrightarrow \begin{cases} n' \text{ generalized moment constraints on } \mu, & \mathbb{E}_\mu[\varphi_j^f] \leq 0 \\ n_k \text{ generalized moment constraints on } \mu_k, & \mathbb{E}_{\mu_k}[\psi_{k,j}^f] \leq 0 \end{cases}$$

**Theorem**

$$\sup_{(f, \mu) \in \mathcal{A}} \mathbb{E}_\mu[qf] = \sup_{(f, \mu) \in \mathcal{A}_\Delta} \mathbb{E}_\mu[qf]$$

$$\mathcal{A}_\Delta = \left\{ (f, \mu) \in \mathcal{A} \left| \begin{array}{l} \mu_k \text{ is a sum of at most} \\ n' + n_k + 1 \text{ weighted} \\ \text{Dirac measures on } \mathcal{X}_k \end{array} \right. \right\}$$

$$\sup_{(f, \mu) \in \mathcal{A}} \mathbb{E}_{\mu} [q_f]$$

Non-convex and infinite dimensional optimization problems

Can be considered as a generalization of classical Chebyshev inequalities

### Connection between Chebyshev inequalities and optimization theory

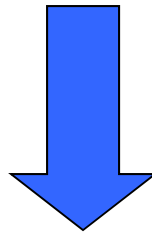
- Classical Markov-Krein theorem (Karlin, Studden, 1958)
- Karlin, Studden (1966, Tchebycheff systems with applications in analysis and statistics)
- Marshall & Olkin (1979, Inequalities: Theory of majorization and its applications)
- Dynkin (1978, Sufficient statistics & extreme points)
- Karr (1983, Extreme points of probability measures with applications)
- Artzner et al (1997, risk measures, value at risk, etc...)
- Betsimas & Popescu (2008, convex optimization approach to inequalities in prob. theo.)

### Our proof rely on a form of Linear programming in infinite dimensional spaces

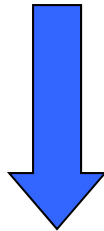
- Winkler (1988, Extreme points of moment sets)
- Follows from an extension of Choquet theory (Phelps 2001, lectures on Choquet's theorem) by Von Weizsacker & Winkler (1979, Integral representation in the set of solutions of a generalized moment problem)
- Kendall (1962, Simplexes & Vector lattices)

# Reduction of optimization variables

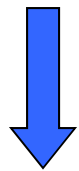
$$\{f: \mathcal{X} \rightarrow \mathbb{R}, \mu \in \mathcal{P}(\mathcal{X})\}$$



$$\left\{ f: \mathcal{X} \rightarrow \mathbb{R}, \mu \in \mathcal{P}(\mathcal{X}) \mid \mu = \sum_{i=1}^k \alpha_k \delta_{x_k} \right\}$$



$$\{f: \{1, 2, \dots, n\} \rightarrow \mathbb{R}, \mu \in \mathcal{P}(\{1, 2, \dots, n\})\}$$



$$\{\{1, 2, \dots, q\}, \mu \in \mathcal{P}(\{1, 2, \dots, n\})\}$$

## Example: Optimal concentration inequality

$$\mathcal{A}_{MD} := \left\{ (f, \mu) \left| \begin{array}{l} f: \mathcal{X}_1 \times \cdots \times \mathcal{X}_m \rightarrow \mathbb{R}, \\ \mu \in \mathcal{M}(\mathcal{X}_1) \otimes \cdots \otimes \mathcal{M}(\mathcal{X}_m), \\ \mathbb{E}_\mu[f] \leq 0, \\ \text{Osc}_i(f) \leq D_i \end{array} \right. \right\}$$

$$\text{Osc}_i(f) := \sup_{(x_1, \dots, x_m) \in \mathcal{X}} \sup_{x'_i \in \mathcal{X}_i} (f(\dots, x_i, \dots) - f(\dots, x'_i, \dots)).$$

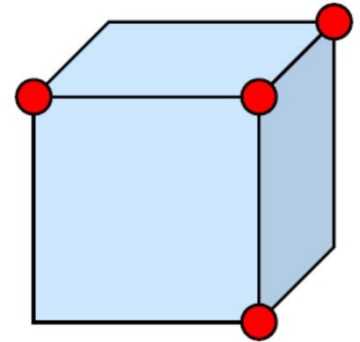
$$\mathcal{U}(\mathcal{A}_{MD}) := \sup_{(f, \mu) \in \mathcal{A}_{MD}} \mu[f(X) \geq a]$$

**McDiarmid inequality**

$$\mathcal{U}(\mathcal{A}_{MD}) \leq \exp\left(-2 \frac{a^2}{\sum_{i=1}^m D_i^2}\right)$$

# Reduction of optimization variables

$$\mathcal{A}_C := \left\{ (C, \alpha) \mid \begin{array}{l} C \subset \{0, 1\}^m, \\ \alpha \in \bigotimes_{i=1}^m \mathcal{M}(\{0, 1\}), \\ \mathbb{E}_\alpha[h^C] \leq 0 \end{array} \right\}$$



$$h^C : \{0, 1\}^m \longrightarrow \mathbb{R}$$

$$t \longrightarrow a - \min_{s \in C} \sum_{i: s_i \neq t_i} D_i$$

$$\mathcal{U}(\mathcal{A}_C) := \sup_{(C, \alpha) \in \mathcal{A}_C} \alpha[h^C \geq a]$$

**Theorem**

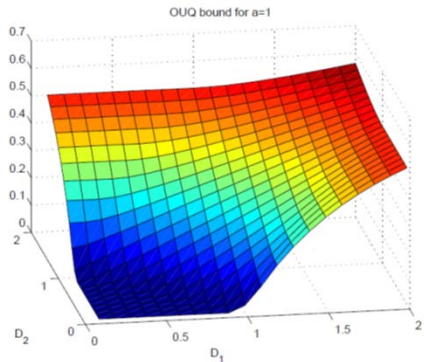
$$\mathcal{U}(\mathcal{A}_{MD}) = \mathcal{U}(\mathcal{A}_C)$$

# Explicit Solution $m=2$

**Theorem**  $m = 2$

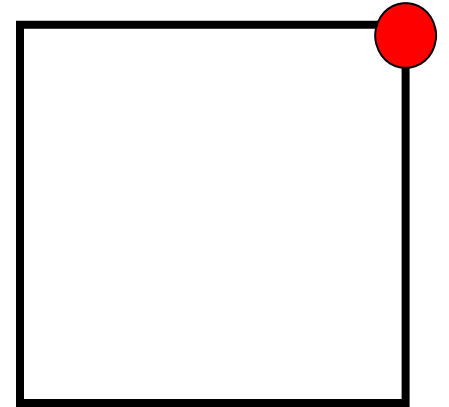
$$\mathcal{U}(\mathcal{A}_{MD}) = \begin{cases} 0 & \text{if } D_1 + D_2 \leq a \\ \frac{(D_1 + D_2 - a)^2}{4D_1 D_2} & \text{if } |D_1 - D_2| \leq a \leq D_1 + D_2 \\ 1 - \frac{a}{\max(D_1, D_2)} & \text{if } 0 \leq a \leq |D_1 - D_2| \end{cases}$$

**OUQ bound  $a=1$**



$$C = \{(1, 1)\}$$

$$h^C(s) = a - (1 - s_1)D_1 - (1 - s_2)D_2$$



**Corollary** If  $D_1 \geq a + D_2$ , then

$$\mathcal{U}(\mathcal{A}_{MD})(a, D_1, D_2) = \mathcal{U}(\mathcal{A}_{MD})(a, D_1, 0)$$

# Reduction calculus with measures over measures

$$\begin{array}{ccc} \mathcal{M}(\mathcal{X}) \supset \mathcal{A} & \xrightarrow{\Psi} & \mathcal{Q} & \text{Polish space} \\ \mathcal{M}(\mathcal{A}) \supset \Pi & \xleftarrow{\Psi^{-1}} & \mathcal{Q} & \subset \mathcal{M}(\mathcal{Q}) \end{array}$$

## Theorem

$$\begin{array}{c} \sup_{\pi \in \Psi^{-1} \mathcal{Q}} \mathbb{E}_{\mu \sim \pi} [\Phi(\mu)] \\ \parallel \\ \sup_{\mathcal{Q} \in \mathcal{Q}} \left[ \mathbb{E}_{q \sim \mathcal{Q}} \left[ \sup_{\mu \in \Psi^{-1}(q)} \Phi(\mu) \right] \right] \end{array}$$



## A simple example

10,000 children are given one pound of play-doh. On average, how much mass can they put above  $a$  While, on average, keeping the seesaw balanced around  $m$ ?



Paul is given one pound of play-doh. What can you say about how much mass he is putting above  $a$  if all you have is the belief that he is keeping the seesaw balanced around  $m$ ?

What is the least upper bound on

$$\mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]]$$

If all you know is  $\mathbb{E}_{\mu \sim \pi} [\mathbb{E}_{\mu}[X]] = m$  ?



$$\mu \in \mathcal{A} := \mathcal{M}([0, 1])$$

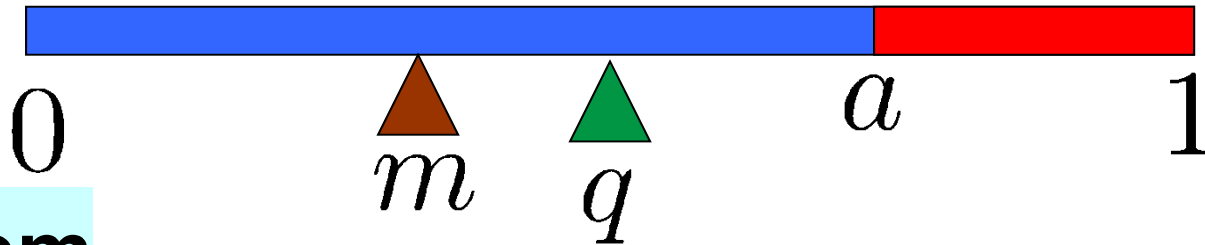
Answer

$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]]$$

$$\Pi := \left\{ \pi \in \mathcal{M}(\mathcal{A}) : \mathbb{E}_{\mu \sim \pi} [\mathbb{E}_{\mu}[X]] = m \right\}$$

$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]]$$

$$\Pi := \left\{ \pi \in \mathcal{M}(\mathcal{M}([0, 1])) : \mathbb{E}_{\mu \sim \pi} [\mathbb{E}_{\mu}[X]] = m \right\}$$



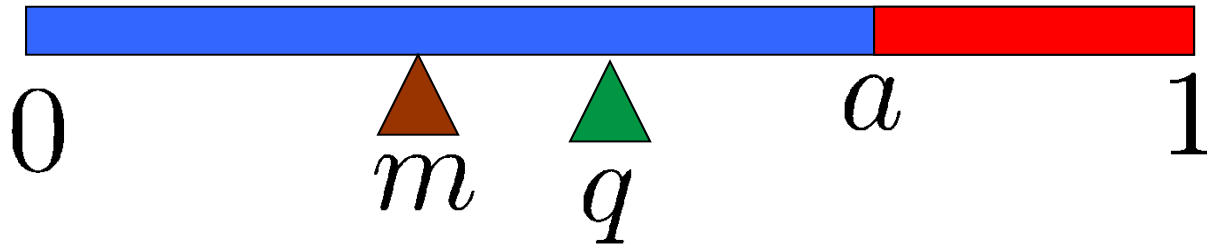
## Theorem

$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]] = \sup_{\mathbb{Q} \in \mathcal{M}([0, 1]) : \mathbb{E}_{\mathbb{Q}}[q] = m}$$

$$\mathbb{E}_{q \sim \mathbb{Q}} \left[ \sup_{\mu \in \mathcal{M}([0, 1]) : \mathbb{E}_{\mu}[X] = q} \mu[X \geq a] \right]$$

$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]]$$

$$\Pi := \left\{ \pi \in \mathcal{M}(\mathcal{M}([0, 1])) : \mathbb{E}_{\mu \sim \pi} [\mathbb{E}_{\mu}[X]] = m \right\}$$



$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]] = \sup_{\mathbb{Q} \in \mathcal{M}([0, 1]) : \mathbb{E}_{\mathbb{Q}}[q] = m} \mathbb{E}_{q \sim \mathbb{Q}} \left[ \min\left(\frac{q}{a}, 1\right) \right]$$

$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]]$$

$$\Pi := \left\{ \pi \in \mathcal{M}(\mathcal{M}([0, 1])) : \mathbb{E}_{\mu \sim \pi} [\mathbb{E}_{\mu}[X]] = m \right\}$$



$$\sup_{\pi \in \Pi} \mathbb{E}_{\mu \sim \pi} [\mu[X \geq a]] = \frac{m}{a}$$

**Can this form of calculus in infinite dimensional spaces facilitate the process of scientific discovery?**

**New Reproducing Kernel Hilbert Spaces and Selberg Integral formulas**

**Forrester and Warnaar 2008**

**The importance of the Selberg integral**

“Used to prove outstanding conjectures in Random matrix theory and cases of the Macdonald conjectures”

“Central role in random matrix theory, Calogero-Sutherland quantum many-body systems, Knizhnik-Zamolodchikov equations, and multivariable orthogonal polynomial theory”

# The truncated moment problem

$$\begin{array}{ccc} \mathcal{M}[0, 1] & \xrightarrow{\Psi} & \mathbb{R}^k \\ \mu & \xrightarrow{\quad} & \left( \mathbb{E}_{X \sim \mu}[X], \mathbb{E}_{X \sim \mu}[X^2], \dots, \mathbb{E}_{X \sim \mu}[X^k] \right) \end{array}$$

Study of the geometry of  $M_k := \Psi(\mathcal{M}([0, 1]))$



**P. L. Chebyshev**  
1821-1894



**A. A. Markov**  
1856-1922



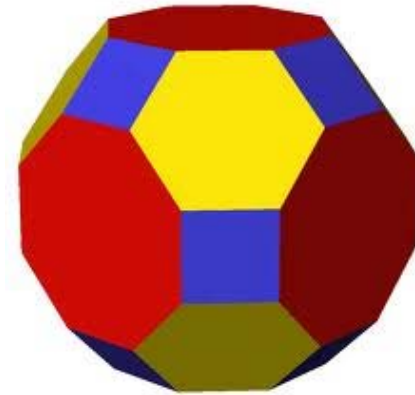
**M. G. Krein**  
1907-1989

$$\mathcal{M}[0, 1] \xrightarrow{\Psi} \mathbb{R}^k$$

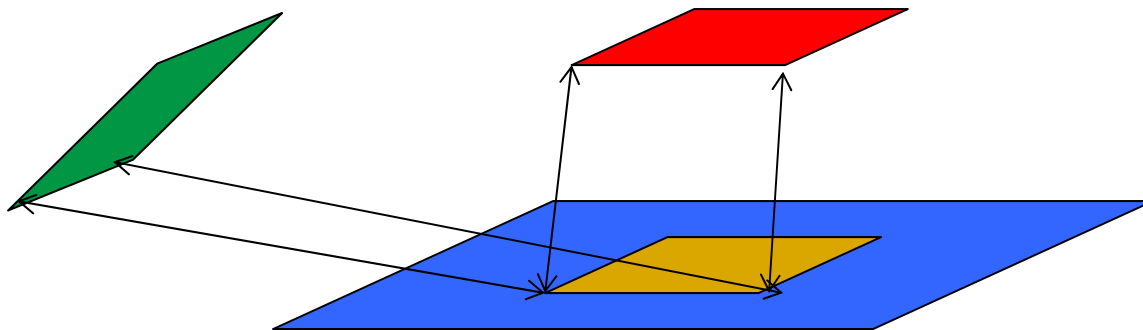
$$\mu \xrightarrow{\quad} \left( \mathbb{E}_{X \sim \mu}[X], \mathbb{E}_{X \sim \mu}[X^2], \dots, \mathbb{E}_{X \sim \mu}[X^k] \right)$$

$$M_k := \Psi(\mathcal{M}([0, 1]))$$

**Origin of these new Selberg  
integral formulas and new RKHS**



Compute  $\text{Vol}(M_k)$  using different  
(finite-dimensional) representations in  $\mathcal{M}([0, 1])$



**Infinite dim.**

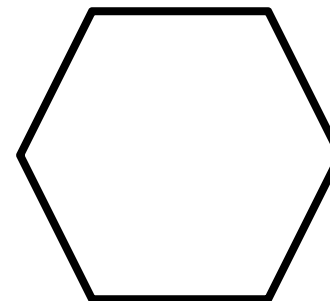
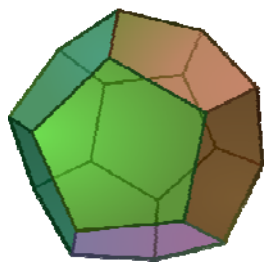


**Finite dim.**



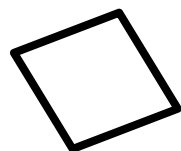
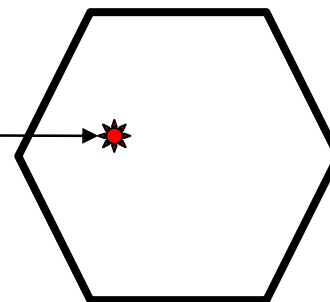
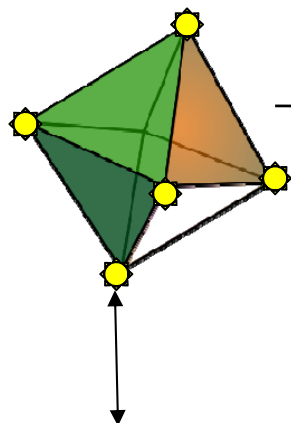
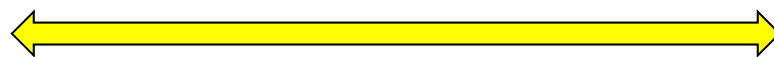
$$\mathcal{M}[0, 1] \xrightarrow{\Psi} \mathbb{R}^k \quad \boxed{M_k := \Psi(\mathcal{M}([0, 1]))}$$

$$\mu \xrightarrow{\quad} \left( \mathbb{E}_{X \sim \mu}[X], \mathbb{E}_{X \sim \mu}[X^2], \dots, \mathbb{E}_{X \sim \mu}[X^k] \right)$$



**Infinite dim.**

**Finite dim.**

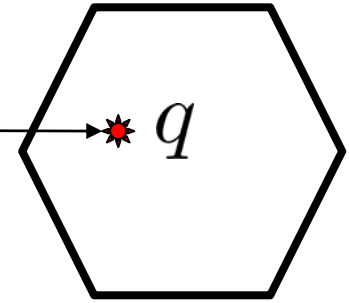
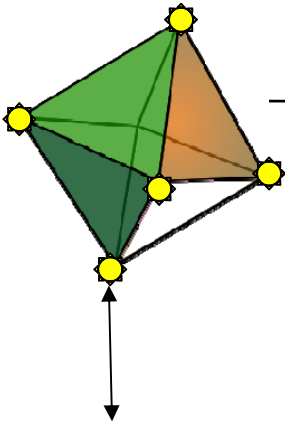


**Finite dim.**

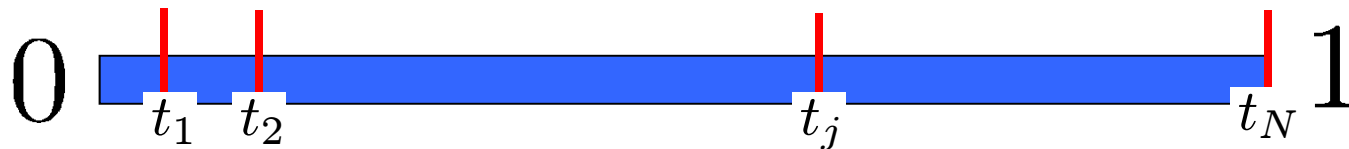
Let us compute  $\text{Vol}(M_k)$  using different extreme points representations.

**Infinite dim.**

**Finite dim.**



$$\mu = \sum_{j=1}^N \lambda_j \delta_{t_j} \xrightarrow{\Psi} (q_1, \dots, q_k)$$
$$q_i = \sum_{j=1}^N \lambda_j t_j^i$$



$$\mu = \sum_{j=1}^N \lambda_j \delta_{t_j}$$

**Index**  $i(\mu)$ : Number of support points of  $\mu$

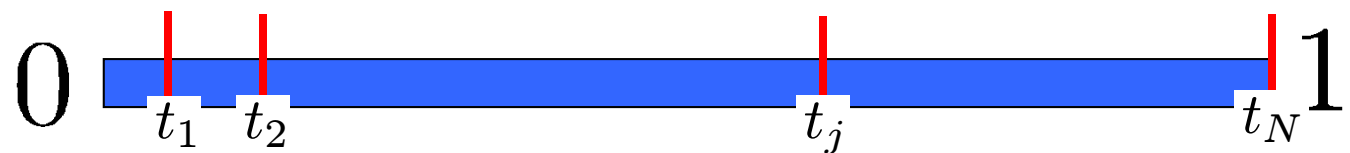
Counting interior points with weight 1 and boundary points with weight  $\frac{1}{2}$

- $\mu$  is called
- principal if  $i(\mu) = \frac{k+1}{2}$
  - canonical if  $i(\mu) = \frac{k+2}{2}$
  - upper if support points include 1
  - lower if support points do not include 1

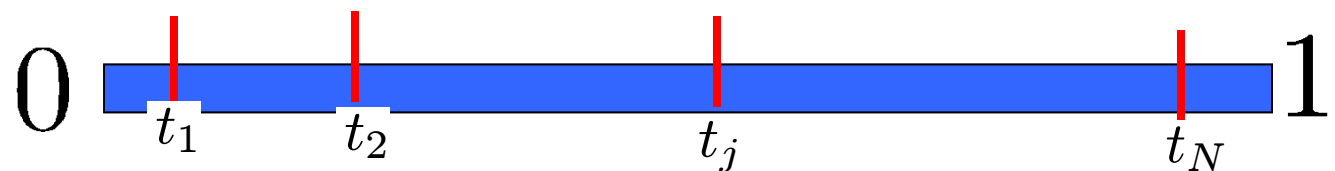
## Theorem

Every point  $q \in \text{Int}(M_k)$  has a unique upper and lower principal representation.

**Upper**



**Lower**



$\text{Vol}(M_{2m-1})$  using Upper Rep. =  $\text{Vol}(M_{2m-1})$  using Lower Rep.

$$\frac{1}{(m-1)!} S_{m-1}(3, 3, 2) = \frac{1}{m!} S_m(1, 1, 2)$$

$\text{Vol}(M_{2m})$  using Upper Rep. =  $\text{Vol}(M_{2m})$  using Lower Rep.

$$S_m(1, 3, 2) = S_m(3, 1, 2)$$

## Selberg Identities

$$S_n(\alpha, \beta, \gamma) = \prod_{j=0}^{n-1} \frac{\Gamma(\alpha+j\gamma)\Gamma(\beta+j\gamma)\Gamma(1+(j+1)\gamma)}{\Gamma(\alpha+\beta+(n+j-1)\gamma)\Gamma(1+\gamma)}$$

$$S_n(\alpha, \beta, \gamma) := \int_{[0,1]^n} \prod_{j=1}^n t_j^{\alpha-1} (1-t_j)^{\beta-1} |\Delta(t)|^{2\gamma} dt.$$

$$\Delta(t) := \prod_{j < k} (t_k - t_j)$$

Brittleness of Bayesian inference and new Selberg formulas. H. Owhadi and C. Scovel. Communications in Mathematical Sciences (2015). arXiv:1304.7046

$$\mu = \sum_{j=1}^N \lambda_j \delta_{t_j}$$

**Index**  $i(\mu)$ : Number of support points of  $\mu$

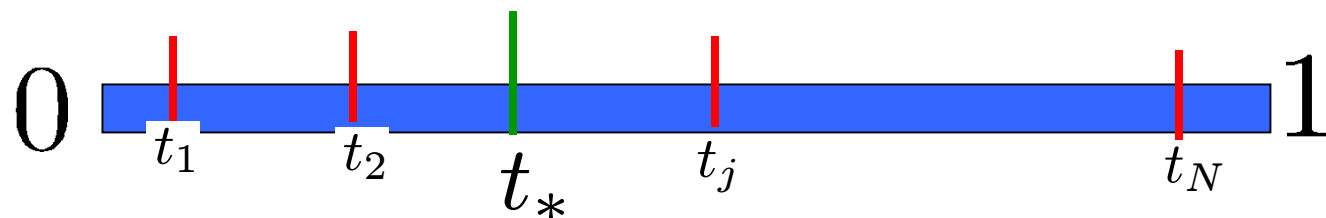
Counting interior points with weight 1 and boundary points with weight  $\frac{1}{2}$

- $\mu$  is called
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  - upper if support points include 1
  - lower if support points do not include 1

## Theorem

For  $t_* \in (0, 1)$ , every point  $q \in \text{Int}(M_k)$  has a unique canonical representation whose support contains  $t_*$ .

When  $t_* = 0$  or 1, there exists a unique principal representation whose support contains  $t_*$ .



**New Reproducing Kernel Hilbert Spaces and Selberg Integral formulas related to the Markov-Krein representations of moment spaces.**

$$\begin{array}{ccc} \mathcal{M}[0, 1] & \xrightarrow{\Psi} & [0, 1]^k \\ \mu & \xrightarrow{\quad} & \left( \mathbb{E}_{X \sim \mu}[X], \mathbb{E}_{X \sim \mu}[X^2], \dots, \mathbb{E}_{X \sim \mu}[X^k] \right) \end{array}$$

$$\int_{I^m} \Sigma t^{-1} \cdot \prod_{j=1}^m t_j^2 (1 - t_j)^2 \Delta_m^4(t) dt = \frac{S_m(5, 1, 2) - S_m(3, 3, 2)}{2}$$

$$\int_{I^m} \Sigma t^{-1} \cdot \prod_{j=1}^m t_j^2 \cdot \Delta_m^4(t) dt = \frac{m}{2} S_{m-1}(5, 3, 2)$$

$$\Delta_m(t) := \prod_{j < k} (t_k - t_j) \quad I := [0, 1]$$

$$(\Sigma \phi)(t) := \sum_{j=1}^m \phi(t_j), \quad t \in I^m$$

$$S_n(\alpha, \beta, \gamma) = \prod_{j=0}^{n-1} \frac{\Gamma(\alpha + j\gamma) \Gamma(\beta + j\gamma) \Gamma(1 + (j+1)\gamma)}{\Gamma(\alpha + \beta + (n+j-1)\gamma) \Gamma(1 + \gamma)}$$

$$e_j(t) := \sum_{i_1 < \dots < i_j} t_{i_1} \cdots t_{i_j}$$

$\Pi_0^n$ :  $n$ -th degree polynomials which vanish on the boundary of  $[0, 1]$

$M_n \subset \mathbb{R}^n$ : set of  $q = (q_1, \dots, q_n) \in \mathbb{R}^n$  such that there exists a probability measure  $\mu$  on  $[0, 1]$  with  $\mathbb{E}_\mu[X^i] = q_i$  with  $i \in \{1, \dots, n\}$ .

## Theorem Bi-orthogonal systems of Selberg Integral formulas

Consider the basis of  $\Pi_0^{2m-1}$  consisting of the associated Legendre polynomials  $Q_j, j = 2, \dots, 2m - 1$  of order 2 translated to the unit interval  $I$ . For  $k = 2, \dots, 2m - 1$  define

$$a_{jk} := \frac{(j + k + k^2)\Gamma(j + 2)\Gamma(j)}{\Gamma(j + k + 2)\Gamma(j - k + 1)}, \quad k \leq j \leq 2m - 1$$

$$\tilde{h}_k(t) := \sum_{j=k}^{2m-1} (-1)^{j+1} a_{jk} e_{2m-1-j}(t, t).$$

Then for  $j = k \pmod{2}, j, k = 2, \dots, 2m - 1$ , we have

$$\int_{I^{m-1}} \tilde{h}_k(t) \Sigma Q_j(t) \prod_{j'=1}^{m-1} t_{j'}^2 \cdot \Delta_{m-1}^4(t) dt = \text{Vol}(M_{2m-1}) (2m-1)! (m-1)! \frac{(k+2)!}{(8k+4)(k-2)!} \delta_{jk}.$$