

# Beyond SICs and MUBs: The conical designs

Anna Szymusiak <sup>1</sup>

*in collaboration with:*

Michele Dall'Arno<sup>2</sup>, Sarah Brandsen<sup>2,3</sup>, Jung Jun Park<sup>2</sup>

<sup>1</sup>Institute of Mathematics, Jagiellonian University

<sup>2</sup>Centre for Quantum Technologies, National University of Singapore

<sup>3</sup>California Institute of Technology

# Outline

Spherical and projective  $t$ -designs

Conical 2-designs

Informational power of POVM

Informational power of some conical designs

## Definition

A **spherical  $t$ -design** is a set of  $n$  normalized vectors  $\{\phi_k\}_{k=1}^n \subset \mathcal{S}^d$  such that the average value of any  $t$ -th order polynomial  $f_t$  over the set  $\{\phi_k\}_{k=1}^n$  is equal to the average of  $f_t$  over all normalized vectors:

$$\frac{1}{n} \sum_{k=1}^n f_t(\phi_k) = \int_{\mathcal{S}^d} f_t(\phi) d\phi.$$

## Example ( $d = 3$ )

- ▶ 1-design: triangular bipyramid
- ▶ 2-design: tetrahedron
- ▶ 3-design: cube, octahedron
- ▶ 5-design: icosahedron, dodecahedron

## Definition

A **complex projective  $t$ -design** is a set of  $n$  pure states  $\{\rho_k\}_{k=1}^n \subset \mathcal{P}(\mathbb{C}^d)$  such that

$$\frac{1}{n^2} \sum_{k,l=1}^n f_t(\text{tr}(\rho_k \rho_l)) = \iint_{\mathcal{P}(\mathbb{C}^d)^2} f_t(\text{tr} \rho \sigma) d\mu_H(\rho) d\mu_H(\sigma)$$

for any any  $t$ -th order polynomial  $f_t$ , where  $\mu_H$  denotes the unique unitarily invariant probability measure on  $\mathcal{P}(\mathbb{C}^d)$ .

Equivalently:

$$\frac{1}{n} \sum_{k=1}^n \rho_k^{\otimes t} = \binom{d+t-1}{t}^{-1} P_s,$$

where  $P_s$  is the projection onto symmetric subspace of  $(\mathbb{C}^d)^{\otimes t}$ .

**Note:** Every rank-1 normalized POVM can be considered as a complex projective 1-design

## Examples of complex projective 2-designs

### SIC-POVMs

A **symmetric informationally complete POVM (SIC-POVM)** consists of  $d^2$  subnormalized rank-one projections  $\Pi_j = |\phi_j\rangle\langle\phi_j|/d$  such that:

$$\text{tr}(\Pi_i^* \Pi_j) = \frac{|\langle\phi_i|\phi_j\rangle|^2}{d^2} = \frac{1}{d^2(d+1)} \text{ for } i \neq j.$$

The states  $\{|\phi_j\rangle\langle\phi_j|\}_{j=1}^{d^2}$  constitute a complex projective 2-design.

### MUBs

The complete set of **mutually unbiased bases (MUB)** is a set of  $d+1$  orthonormal bases  $\{e_i^j\}_{i=1, j=1}^{d, d+1}$  in  $\mathbb{C}^d$  such that

$$|\langle e_i^j | e_k^l \rangle|^2 = 1/d, \quad \text{for } j \neq l.$$

The states  $\{|e_i^j\rangle\langle e_i^j|\}_{i=1, j=1}^{d, d+1}$  constitute a complex projective 2-design.

## Theorem (Graydon, Appleby (2016))

Let  $\{A_1, \dots, A_m\}$  be a family of positive-semidefinite operators on  $\mathbb{C}^d$ . Then the following statements are equivalent:

- i.  $\sum_{j=1}^m A_j \otimes A_j$  commutes with  $U \otimes U$  for every unitary  $U$ .
- ii. For some  $k_s \geq k_a \geq 0$

$$\sum_{j=1}^m A_j \otimes A_j = k_s P_s + k_a P_a.$$

Moreover, the  $A_j$  span the space of selfadjoint operators on  $\mathbb{C}^d$  iff  $k_s > k_a$ .

### Definition

A conical 2-design is a family of non-zero positive-semidefinite operators  $\{A_1, \dots, A_m\}$  on  $\mathbb{C}^d$  satisfying the equivalent conditions in the above theorem with  $k_s > k_a$ .

M.A. Graydon, D.M. Appleby, J. Phys.A **49**, 085301 (2016)

Let  $t_j := \text{tr}A_j$ , and  $t := \sqrt{\frac{1}{m} \sum_j t_j^2}$ . Then

$$\sum_j t_j A_j = \frac{mt^2}{d} I$$

and the operators

$$E_j = \frac{dt_j}{mt^2} A_j$$

constitute a POVM. If  $t_j = \text{const}$  this POVM is also a conical 2-design.

### Lemma (Index of coincidence)

*The index of coincidence equals  $k_s$ , i.e.*

$$\sum_j \text{tr}^2[A_j \rho] = k_s$$

*for every pure state  $\rho$ .*

## Some classes of conical 2-designs

- ▶ weighted projective 2-designs:

$$\sum_{k=1}^n w(k) \rho_k^{\otimes t} = \binom{d+t-1}{t}^{-1} P_s, \quad \text{where } \sum_k w(k) = 1$$

- ▶ homogeneous conical 2-designs ( $\text{tr} A_j = \text{const}$  and  $\text{tr}(A_j^2) = \text{const}$ )
  - ▶ SIMs and MUMs (arbitrary rank generalization of SICs and MUBs)
    - ▶ **anti-SICs and anti-MUBs**: POVMs consisting of subnormalized projectors onto  $d - 1$ -dimensional subspaces orthogonal to the states defining original POVM
    - ▶ **depolarized SICs, MUBs and their anti-counterparts**: convex mixtures of original POVM elements with identity



## Lemma

For any conical 2 design  $A_1, \dots, A_m$  let  $a_j$  and  $b_j$  be the maximum and minimum eigenvalues of operator  $A_j$ , respectively. Then the affine combination  $A_{j,\lambda} := \lambda A_j + (1 - \lambda) \frac{1}{m} \mathbb{I}$  with

$$\lambda \in \left[ \max_j \frac{1}{1 - ma_j}, \min_j \frac{1}{1 - mb_j} \right],$$

is also a conical 2-design.

In particular, for rank-1  $A_j$ ,  $\frac{t_j \mathbb{I} - A_j}{d-1}$  (the anti-  $A_j$ ) is a conical 2-design.

$V := \{\pi_i, \rho_i\}_{i=1}^l$  – an ensemble of initial states,  $\Pi = \{\Pi_j\}_{j=1}^k$  – a POVM

*How much information can be extracted from  $V$  by measurement  $\Pi$ ?*

### Definition

The **mutual information between  $V$  and  $\Pi$**  is given by

$$I(V, \Pi) := \sum_{i=1}^l \eta \left( \sum_{j=1}^k P_{ij} \right) + \sum_{j=1}^k \eta \left( \sum_{i=1}^l P_{ij} \right) - \sum_{i=1}^l \sum_{j=1}^k \eta(P_{ij}),$$

where  $P_{ij} = \pi_i \text{tr}(\rho_i \Pi_j)$  for  $i = 1, \dots, l$  and  $j = 1, \dots, k$ , and  $\eta(x) := -x \ln x$  ( $x > 0$ ),  $\eta(0) = 0$ .

*What is the capability of extracting information by given measurement?*

### Definition

The **informational power** of  $\Pi$  is denoted by  $W(\Pi)$  and given by

$$W(\Pi) := \max_{V-\text{ensemble}} I(V, \Pi).$$

M. Dall'Arno, G.M. D'Ariano, and M.F. Sacchi, Phys. Rev. A **83**, 062304 (2011)

O. Oreshkov, J. Calsamiglia, R. Muñoz Tapia, and E. Bagan, New J. Phys. **13**, 073032 (2011)



- ▶ There exists a *maximally informative ensemble* consisting of pure states only
- ▶ The informational power and the *Shannon entropy of measurement*  $\Pi$ , defined for an initial state  $\rho$  by  $H(\rho, \Pi) := \sum_{j=1}^k \eta(\text{tr}(\rho \Pi_j))$  are related by

$$W(\Pi) \leq \ln k - \min_{\rho} H(\rho, \Pi)$$

with equality iff there exists an ensemble  $V = \{\pi_i, \rho_i\}_{i=1}^l$  such that the states  $\rho_i$  are minimizers of  $H(\cdot, \Pi)$  and  $\text{tr}((\sum_{i=1}^l \pi_i \rho_i) \Pi_j) = 1/k$  for every  $j$

- ▶ The cases in which the informational power has been computed analytically so far:
  - ▶ all highly symmetric POVMs in dimension 2: seven sporadic measurements<sup>3</sup>, including the 'tetrahedral' SIC-POVM<sup>1,2</sup>, and one infinite series<sup>2,3</sup>,
  - ▶ all SIC-POVMs in dimension three<sup>4</sup>,
  - ▶ the POVM consisting of four MUBs in dimension three<sup>5</sup>
  - ▶ the Hoggar SIC-POVM in dimension eight<sup>6</sup>

<sup>1</sup>M. Dall'Arno, G.M. D'Ariano, and M.F. Sacchi, Phys. Rev. A **83**, 062304 (2011)

<sup>2</sup>O. Oreshkov, J. Calsamiglia, R. Muñoz Tapia, and E. Bagan, New J. Phys. **13**, 073032 (2011)

<sup>3</sup>W. Słomczyński and AS, arXiv:1402.0375[quant-ph], to appear in Quantum Inf. Process. (2016)

<sup>4</sup>AS, J. Phys. A **47**, 445301 (2014)

<sup>5</sup>M. Dall'Arno, Phys. Rev. A **90**, 052311 (2014)

<sup>6</sup>W. Słomczyński and AS, arXiv:1512.01735

## The optimization method based on the Hermite interpolation

- ▶ a sequence of points:  $a \leq t_1 < t_2 < \dots < t_m \leq b$
- ▶ a sequence of positive integers:  $k_1, k_2, \dots, k_m$
- ▶ a real valued function  $f \in C^D([a, b])$ , where  $D := k_1 + k_2 + \dots + k_m$

There exists a unique polynomial  $p$  of degree less than  $D$  that agree with  $f$  at  $t_i$  up to a derivative of order  $k_i - 1$  (for  $1 \leq i \leq m$ ), that is,

$$p^{(k)}(t_i) = f^{(k)}(t_i), \quad 0 \leq k < k_i. \quad (1)$$

We will need the following formula for the error in Hermite interpolation:

### Lemma

For each  $t \in (a, b)$  there exists  $\xi \in (a, b)$  such that  $\min\{t, t_1\} < \xi < \max\{t, t_m\}$  and

$$f(t) - p(t) = \frac{f^{(D)}(\xi)}{D!} \prod_{i=1}^m (t - t_i)^{k_i}. \quad (2)$$

Assume that all the derivatives of  $f$  of even order are strictly negative in  $(a, b)$  and these of odd order greater than 1 are strictly positive (holds true for  $\eta$ ):

$$f^{2l}(x) < 0, \quad f^{2l+1}(x) > 0 \quad \text{for } x \in (a, b), \quad l = 1, 2, \dots \quad (3)$$

Moreover, let us assume that

$$k_i := \begin{cases} 1, & \text{if } t_i \in \{a, b\} \\ 2, & \text{otherwise} \end{cases}. \quad (4)$$

## Observation

*Under above assumptions, the Hermite polynomial  $p$  interpolates  $f$*

1. *from below, if  $t_1 = a$ ,*
2. *from above, if  $t_1 > a$ .*

*Moreover,  $f(t) = p(t)$  if and only if  $t = t_i$  for some  $i = 1, \dots, m$ .*

## Theorem (Informational power of depolarized anti-SICs)

Let  $A_j$  be a  $d$ -dimensional anti-SIC and  $A_{j,\lambda} := [\lambda A_j + (1 - \lambda)\mathbb{I}/d]$  be its depolarized counterpart. Then the informational power of  $A_{j,\lambda}$  is given by

$$\ln d + \frac{1}{d}\eta\left(\frac{1-\lambda}{d}\right) + \frac{d^2-1}{d}\eta\left(\frac{d^2-1+\lambda}{d(d^2-1)}\right)$$

and the maximally informative ensemble consists of the corresponding SIC ensemble.

### Sketch of the proof.

The optimal probability distribution consists of the smallest achievable and  $d^2 - 1$  equal other probabilities. We apply the Hermite interpolation method. □

## Theorem (Informational power of depolarized qubit SICs)

The informational power of any depolarized 2-dimensional SIC  $A_{j,\lambda}$  is given by

$$\ln 2 + \frac{1}{2}\eta \left( \frac{1-\lambda}{2} \right) + \frac{3}{2}\eta \left( \frac{3+\lambda}{6} \right)$$

and the maximally informative ensemble consists of the corresponding anti-SIC ensemble.

### Sketch of the proof.

The optimal probability distribution consists of the smallest achievable and 3 equal other probabilities. We apply the Hermite interpolation method.  $\square$

## Theorem (Informational power of depolarized qubit complete MUBs)

*The informational power of any depolarized 2-dimensional complete MUB  $A_{j,\lambda}$  is given by*

$$\ln(2) + \frac{1}{3}\eta\left(\frac{1-\lambda}{2}\right) + \frac{4}{3}\eta\left(\frac{1}{2}\right) + \frac{1}{3}\eta\left(\frac{1+\lambda}{2}\right)$$

*and the maximally informative ensemble consists of the corresponding MUB ensemble.*

### Sketch of the proof.

The optimal probability distribution consists of the smallest achievable, the largest achievable and 4 equal other probabilities. We apply the Hermite interpolation method. □



## Theorem (Informational power of depolarized qutrit SICs)

*The informational power of any depolarized 3-dimensional SIC  $A_{j,\lambda}$  is given by*

$$\ln(3) + \eta\left(\frac{1-\lambda}{3}\right) + 2\eta\left(\frac{2+\lambda}{6}\right)$$

*and the maximally informative ensemble is an orthonormal basis consisting of elements orthogonal to exactly 3 elements of original SIC.*

### Sketch of the proof.

The optimal probability distribution consists of 3 the smallest achievable, and 6 equal other probabilities. We apply the Hermite interpolation method.  $\square$

## Theorem (Informational power of depolarized qutrit complete MUBs)

*The informational power of any depolarized 3-dimensional complete MUB  $A_{j,\lambda}$  is given by*

$$\ln(3) + \eta\left(\frac{1-\lambda}{3}\right) + 2\eta\left(\frac{2+\lambda}{6}\right)$$

*and the maximally informative ensemble consists of the elements of corresponding Hesse SIC.*

### Sketch of the proof.

The optimal probability distribution consists of 4 the smallest achievable, and 8 equal other probabilities. We apply the Hermite interpolation method.  $\square$

## Theorem (Informational power of depolarized Hoggar SICs)

*The informational power of any depolarized Hoggar SIC  $A_{j,\lambda}$  is given by*

$$\ln 8 + \frac{7}{2}\eta\left(\frac{1-\lambda}{8}\right) + \frac{9}{2}\eta\left(\frac{9+7\lambda}{72}\right)$$

*and the maximally informative ensemble consists of the elements of the dual Hoggar SIC.*

### Sketch of the proof.

The optimal probability distribution consists of 28 the smallest achievable, and 32 equal other probabilities. We apply the Hermite interpolation method.  $\square$