

# Optimal universal learning machines for quantum state discrimination

Marco Fanizza, Andrea Mari, Vittorio Giovannetti

**Abstract**—We consider the problem of correctly classifying a given quantum two-level system (qubit) which is known to be in one of two equally probable quantum states. We assume that this task should be performed by a quantum machine which does not have at its disposal a complete classical description of the two template states, but can only have partial prior information about their level of purity and mutual orthogonality. Moreover, similarly to the classical supervised learning paradigm, we assume that the machine can be trained by  $n$  qubits prepared in the first template state and by  $n$  more qubits prepared in the second template state. In this situation we are interested in the optimal process which correctly classifies the input qubit with the largest probability allowed by quantum mechanics. The problem is studied in its full generality for a number of different prior information scenarios and for an arbitrary size  $n$  of the training data. Finite size corrections around the asymptotic limit  $n \rightarrow \infty$  are also derived.

## I. INTRODUCTION

Machine Learning (ML) is that branch of computer science which studies how to instruct a computer to solve a specific task by feeding it with a collection of training data from which it could learn how to proceed. This approach finds applications in a variety of practical pattern recognition, decision and clustering problems where a definite classification of the various alternatives are not directly accessible [1]. Not surprisingly, the interplay between ML and quantum information is very promising (see [2], [3], [4], [5], [6] and references therein). ML has been proposed as a useful tool to improve the

performances of a variety of quantum information procedures, e.g. identification of optimal quantum measurement and estimation procedures, quantum gate design and quantum dynamics engineering. On the other hand, it has been shown that quantum computing can provide speed-ups for ML problems. Moreover, as originally hinted in Refs. [8], [9], [10], [11], [12], a drastic departure from classical data analysis is instead realized in Quantum Learning (QL), where the “learning from examples” paradigm is adopted as a new mode of operation of quantum devices which have access to (not necessarily classical) training data. This setting appears to be perfectly suited to deal with the specific character of quantum mechanics where, at variance with classical models, a fundamental discrepancy exists between the state of a system and the “knowledge” one can acquire about it through measurements. Such discrepancy is a distinctive feature of the theory: ultimately it can be traced back to the no-cloning theorem [16] and poses intrinsic limitations on information retrieval processes. Accordingly in quantum mechanics, the ability of perfectly discriminating alternative configurations, let them being states or processes, can only be guaranteed under special conditions (semiclassical limit). Since the seminal works of Helstrom [13], Holevo [14] and Yuen *et al.* [15], developing optimal probabilistic strategies to face these limitations is a fundamental problem of quantum information. A standard example is provided by quantum state discrimination: here an agent is presented with a quantum system  $Q$  and asked to identify its state knowing that the latter was randomly drawn from an ensemble of possible alternatives which are specified in terms of classical data that fully characterize them. The QL version of this problem is obtained by replacing such classical

Marco Fanizza is with NEST, Scuola Normale Superiore, Istituto Nanoscienze-CNR, I-56126 Pisa, Italy.

Andrea Mari is with NEST, Scuola Normale Superiore, Istituto Nanoscienze-CNR, I-56126 Pisa, Italy.

Vittorio Giovannetti is with NEST, Scuola Normale Superiore and Istituto Nanoscienze-CNR, I-56126 Pisa, Italy.

descriptions with a collection of quantum ancillary systems initialized into the same template states the agent has to assign to  $Q$ . A universal machine for optimal discrimination is hence identified as the quantum device which, by having full access to  $Q$  and the ancillas, allows the agent to solve the identification task with the smallest probability of error. The problem has been addressed in various scenarios in [10], [11], [17], [18], [19], [20], [22], [21], and has attracted the attention of the community as an example of a genuine supervised QL task [4], [5], [7]. In this article we present results about universal machines for qubit discrimination, which can discriminate among any two states, extending in particular the results of [19] to include a variety of scenarios. Specifically we focus on hybrid QL configurations where the agent, beside being provided with the quantum ancillas, has also access to some prior classical information on the templates configurations, such as their purity or their mutual distance. These scenarios naturally emerge when, for instance, the training and the target data are effected by some deteriorating processes (say dephasing or decoherence transformations) which the agent cannot prevent from occurring, but whose operating mechanisms are known to him, or when the different templates are affected by uncertainties arising from the absence of a common, shared reference frame.

The manuscript is organized as follows: notation and model are introduced in Sec. II. The principal part of the paper is Sec. III where our results are explicitly derived in three dedicated subsections. In the first two paragraphs we extend the results of the work of Ref. [19]. Specifically in Sec. III-A we study the case of an optimal universal machines which is trained to discriminate between two qubit density matrices of fixed but different purities. In Sec. III-B instead we focus on the case where the training data are two generic (possibly) mixed quantum systems. Finally in Sec. III-C we discuss the scenario where the training data are pure with fixed relative overlap, but otherwise unknown. The interest in this last configuration arises when considering QML processes where, in analogy with the schemes analyzed in Refs. [23], [24], [25], [27], [26], the

party who is creating the template states does not share a common reference frame with the party that is supposed to solve the identification problem. The paper ends with conclusions in Sec. V. Technical material is presented in dedicated appendices.

## II. THE MODEL

In the state discrimination scenario we have in mind a qubit system  $X$  is initialized with probability  $1/2$  in one of two possible template states  $\rho_1$  and  $\rho_2$ . Without having access to the full classical description of these templates configurations (i.e. without knowing the explicit values of their associated Bloch vectors  $\mathbf{r}_1$  and  $\mathbf{r}_2$ , see below), an external agent is now asked to identify which of the two alternative actually occurred by only granting him access to  $X$  and to two independent sets of  $n$  ancillary qubits  $A$  and  $B$ , initialized respectively into  $n$  copies of  $\rho_1$  and  $\rho_2$ . Following Refs. [10], [11], [19], [20] solutions to this problem can be assigned in terms of a two-outcome POVM  $\hat{\mathcal{M}} \equiv \{\hat{\Pi}_1, \hat{\Pi}_2\}$  that acts globally on the full system  $AXB$  formed by the test qubit  $X$  and by the two  $n$ -qubits ancillas. In particular, noticing that the possible states of  $AXB$  are the density matrices  $\tau_1 = \rho_1^{\otimes n} \otimes \rho_1 \otimes \rho_2^{\otimes n}$  (corresponding to have  $X$  in  $\rho_1$ ) and  $\tau_2 = \rho_1^{\otimes n} \otimes \rho_2 \otimes \rho_2^{\otimes n}$  ( $X$  in  $\rho_2$ ), the average error probability of the procedure can be computed as

$$P_{err}^{(n)} = \int d\mu(\rho_1, \rho_2) \frac{\text{Tr}[\tau_1 \hat{\Pi}_2] + \text{Tr}[\tau_2 \hat{\Pi}_1]}{2}, \quad (1)$$

where  $\text{Tr}[\tau_j \hat{\Pi}_j]$  is the probability of success in the identifying the  $j$ -th configuration, while  $d\mu(\rho_1, \rho_2)$  is a measure that gauges the initial ignorance of the agent about  $\rho_1$  and  $\rho_2$ . Exploiting then the completeness relation of  $\hat{\mathcal{M}}$  this can finally recast into

$$P_{err}^{(n)} = \frac{1}{2} - \frac{1}{4} \text{Tr}[\Theta(\hat{\Pi}_1 - \hat{\Pi}_2)], \quad (2)$$

where  $\Theta$  is the trace-null, Hermitian operator

$$\Theta = \alpha^{(n)} - \beta^{(n)}, \quad (3)$$

given by the difference between the following density matrices of  $AXB$ ,

$$\begin{aligned}\alpha^{(n)} &\equiv \int d\mu(\rho_1, \rho_2) \rho_1^{\otimes n} \otimes \rho_1 \otimes \rho_2^{\otimes n}, \\ \beta^{(n)} &\equiv \int d\mu(\rho_1, \rho_2) \rho_1^{\otimes n} \otimes \rho_2 \otimes \rho_2^{\otimes n}.\end{aligned}\quad (4)$$

The minimum (2) can now be easily obtained by choosing as optimal POVM  $\hat{\mathcal{M}}$  which has components  $\hat{\Pi}_1, \hat{\Pi}_2$  respectively projecting on the positive and the negative eigenspaces of  $\Theta$ , i.e.

$$P_{err,min}^{(n)} = \frac{1}{2} - \frac{1}{4} \|\Theta\|_1, \quad (5)$$

with the symbol  $\|\cdot\|_1$  indicating the trace norm.

Some general properties of  $P_{err,min}^{(n)}$  can be determined by simple arguments. First of all since the agent can always discard part of the ancillary states before attempting to identify  $Q$ , for all possible choices of the measure  $d\mu(\rho_1, \rho_2)$ ,  $P_{err,min}^{(n)}$  has to fulfil the inequality

$$P_{err,min}^{(n)} \leq \frac{1}{2} - \frac{1}{4} \left\| \int d\mu(\rho_1, \rho_2) (\rho_1 - \rho_2) \right\|_1, \quad (6)$$

and being a decreasing function of  $n$ , i.e.

$$P_{err,min}^{(n)} \geq P_{err,min}^{(n+1)}. \quad (7)$$

Furthermore, by exploiting the joint-convexity of the trace-norm [28] the following lower bound can be established

$$P_{err,min}^{(n)} \geq \frac{1}{2} - \frac{1}{4} \int d\mu(\rho_1, \rho_2) \|\rho_1 - \rho_2\|_1, \quad (8)$$

for all  $n$  integers. The term on the right-hand-side of this inequality corresponds to the average Helstrom error probability  $\bar{P}_H$ , i.e. the average minimum error probability the agent could attain by providing him/her with a full classical description of the template states: under this condition in fact, for each couple of density matrices  $\rho_1$  and  $\rho_2$ , he/she can tailor a specific POVM on  $X$  that it is optimized to distinguish them. Invoking a full tomographic reconstruction of  $\rho_1$  and  $\rho_2$ , the gap between  $P_{err,min}^{(n)}$  and  $\bar{P}_H$  (optimal excess risk

function [19], [20]), can be shown to nullify in the asymptotic regime  $n \rightarrow \infty$ , i.e.

$$\lim_{n \rightarrow \infty} P_{err,min}^{(n)} = \frac{1}{2} - \frac{1}{4} \int d\mu(\rho_1, \rho_2) \|\rho_1 - \rho_2\|_1 \quad (9)$$

Apart from the above results explicit expressions for  $P_{err,min}^{(n)}$  are known only for a limited set of configurations. For instance in Ref. [10] the Authors focus on the case where both  $\rho_1$  and  $\rho_2$  are pure in general finite dimension, while Ref. [19] provides the formal solution under the assumption that  $\rho_1$  and  $\rho_2$  are density matrices having the same assigned purity. The aim of the present work is to extend these results by expanding the set of treatable scenarios to include the following cases

- i)  $\rho_1$  and  $\rho_2$  having different assigned purities but being otherwise arbitrary;
- ii)  $\rho_1$  and  $\rho_2$  being completely arbitrary (not necessarily pure) density matrices;
- iii)  $\rho_1$  and  $\rho_2$  being arbitrary pure states having assigned mutual distance.

For these configurations we compute the associated values of  $P_{err,min}^{(n)}$  reporting closed analytical expressions for the higher order contributions of their asymptotic expansions at large  $n$ .

### III. DERIVATION

The key ingredient for deriving the above results is the evaluation of the eigenvalues  $\{\lambda_\ell\}_\ell$  of the operator (3) which allows us to rewrite (5) as

$$P_{err,min}^{(n)} = \frac{1}{2} \left( 1 - \sum_\ell^+ \lambda_\ell \right), \quad (10)$$

the sum being restricted on the positive part of the spectrum. For this purpose we exploit the invariance of the average density matrices  $\alpha^{(n)}$  and  $\beta^{(n)}$  under rotations, which effectively transforms the problem in the diagonalization of  $2 \times 2$  matrices, as it was already shown in [19]. Indeed, using the decomposition into irreducible representations of  $SU(2)$ , we notice that the Hilbert space of the  $n$ -qubit systems  $A$  and  $B$  can be expressed as

$$\mathcal{H}_A = \oplus_s \mathcal{H}_A^{(s)}, \quad \mathcal{H}_B = \oplus_t \mathcal{H}_B^{(t)}, \quad (11)$$

where the labels  $s, t$  vary from 0 (if  $n$  is even) or from  $1/2$  (if  $n$  is odd) to  $n/2$ , while  $\mathcal{H}_A^{(s)} = \oplus_i \mathcal{H}_{A_i}^{(s)}$  and  $\mathcal{H}_B^{(t)} = \oplus_k \mathcal{H}_{B_k}^{(t)}$ , respectively, are the direct sums of the irreducible representations with Casimirs  $s(s+1)$  and  $t(t+1)$ , the indexes  $i$  and  $k$  resolving their associated multiplicities – see Eq. (17) below. Accordingly we can then express the joint Hilbert space  $\mathcal{H}_A \otimes \mathcal{H}_X \otimes \mathcal{H}_B$  of our  $2n+1$  qubits system  $AXB$  as the direct sum over  $s, t, i$  and  $k$ , of the following contributions

$$\mathcal{H}_{A_i X B_k}^{(s,t)} = \mathcal{H}_{A_i}^{(s)} \otimes \mathcal{H}_X \otimes \mathcal{H}_{B_k}^{(t)}. \quad (12)$$

As we shall see, in all the three cases detailed in the previous section the associated operator  $\Theta$  is block-diagonal when decomposing  $\mathcal{H}_A \otimes \mathcal{H}_X \otimes \mathcal{H}_B$  in terms of the subspaces  $\mathcal{H}_{A_i X B_k}^{(s,t)}$ . Besides, due to invariance under global rotations on  $AXB$ , for each assigned value of  $s, t, i$ , and  $k$ ,  $\Theta$  further decomposes in a collection of  $2 \times 2$  or  $1 \times 1$  block diagonal matrices whose elements, due to the Wigner-Eckart theorem, only exhibit functional dependence on the indexes  $s, t$  and on the main  $AXB$ -total angular momentum quantum number  $q$  (i.e. the quantum number associated with the full collections of our  $2n+1$  spins and spanning from  $|t-s|-1/2$  to  $t+s+1/2$ ). In particular, by first merging  $A$  and  $X$  and then coupling the two with  $B$ , a convenient orthonormal basis of  $\mathcal{H}_{A_i X B_k}^{(s,t)}$  is provided by the following list of vectors

$$\{|s' = s \pm 1/2, t; q, m\rangle_{i,k}\}_{q,m} \quad (13)$$

where  $m$ , running from  $q$  to  $-q$ , is the associated secondary  $AXB$ -total angular momentum. We stress that in the above construction, and in the remaining of the paper, it is implicit assumed that  $|s \pm 1/2, t; q, m\rangle_{i,k}$  is null whenever the parameters  $s, t$  and  $q$  do not fit the necessary angular momentum selection rules. This allows us to identify four different scenarios:

- a)  $q = s + t + \frac{1}{2}$ ;
- b)  $q = t - s - \frac{1}{2}$  and  $t > s$ ;
- c)  $q = s - t - \frac{1}{2}$  and  $s > t$ ;
- d) all  $s, t, q$  fitting the selection rules which are not included in the previous cases.

In the first three cases, only one of the elements of the couple  $\{|s \pm 1/2, t; q, m\rangle_{i,k}\}$  survives: specifi-

cally the  $s + 1/2$  element for  $a$ ) and  $b$ ), while the  $s - 1/2$  element for  $c$ ). Under such circumstances the symmetry of  $\Theta$  forces it to be  $1 \times 1$  block diagonal, i.e. to admit the associated basis elements as explicit eigenvectors with eigenvalues  $\lambda_{s,t,q}^{(n)}$  that we can formally compute as

$$\Theta_{++}^{(s,t,q)} = {}_{i,k} \langle s + 1/2, t; q, m | \Theta | s + 1/2, t; q, m \rangle_{i,k}, \quad (14)$$

for the case cases  $a$ ) and  $b$ ), and

$$\Theta_{--}^{(s,t,q)} = {}_{i,k} \langle s - 1/2, t; q, m | \Theta | s - 1/2, t; q, m \rangle_{i,k}, \quad (15)$$

for the  $c$ ) case. The corresponding multiplicity is determined instead by the allowed ranges of  $m, i$  and  $k$ , i.e.

$$M_{s,t,q}^{(n)} = (2q+1) \#(s, n) \#(t, n), \quad (16)$$

with  $(2q+1)$  enumerating the possible values of  $m$ , and with  $\#(j, n)$  representing instead the multiplicity of the representations with Casimir  $j(j+1)$  in the decomposition of  $n$  spins  $1/2$ , i.e.

$$\#(j, n) = \frac{n! (2j+1)}{\left(\frac{n-2j}{2}\right)! \left(\frac{n+2j}{2} + 1\right)!}. \quad (17)$$

In the scenario  $d$ ) instead both the elements of the couple  $\{|s \pm 1/2, t; q, m\rangle_{i,k}\}$  survive and the symmetry of the problem forces  $\Theta$  to be described by  $2 \times 2$  block diagonal terms  $\Theta_{i,k}^{(s,t,q,m)}$  of the form,

$$\Theta_{i,k}^{(s,t,q,m)} \equiv \begin{bmatrix} \Theta_{++}^{(s,t,q)} & \Theta_{+-}^{(s,t,q)} \\ \Theta_{-+}^{(s,t,q)} & \Theta_{--}^{(s,t,q)} \end{bmatrix}, \quad (18)$$

with  $\Theta_{++}^{(s,t,q)}$  and  $\Theta_{--}^{(s,t,q)}$  as in (14) and (15) and with

$$\Theta_{+-}^{(s,t,q)} = [\Theta_{-+}^{(s,t,q)}]^* = {}_{i,k} \langle s + 1/2, t; q, m | \Theta | s - 1/2, t; q, m \rangle_{i,k}. \quad (19)$$

Accordingly we get a further set of eigenvalues identified with the functions

$$\lambda_{s,t,q}^{(n)}(\pm) = \left( \frac{\Theta_{--}^{(s,t,q)} + \Theta_{++}^{(s,t,q)}}{2} \right) \pm \sqrt{\left( \frac{\Theta_{--}^{(s,t,q)} - \Theta_{++}^{(s,t,q)}}{2} \right)^2 + |\Theta_{+-}^{(s,t,q)}|^2}, \quad (20)$$

again characterized by multiplicities  $M_{s,t,q}^{(n)}$  defined as in Eq. (16). The corresponding eigenvectors are instead provided by the superpositions

$$|\psi_{s,t,q,m}^{(\pm)}\rangle_{i,k} = A^{(s,t,q)}(\pm)|s+1/2, t; q, m\rangle_{i,k} + B_{s,t,q}^{(n)}(\pm)|s-1/2, t; q, m\rangle_{i,k}, \quad (21)$$

with amplitudes  $A^{(s,t,q)} = \Theta_{+-}^{(s,t,q)}$  and

$$B_{s,t,q}(\pm) = \left( \frac{\Theta_{--}^{(s,t,q)} - \Theta_{++}^{(s,t,q)}}{2} \right) \pm \sqrt{\left( \frac{\Theta_{--}^{(s,t,q)} - \Theta_{++}^{(s,t,q)}}{2} \right)^2 + |\Theta_{+-}^{(s,t,q)}|^2}, \quad (22)$$

which, for easy of notation we present in a non-normalized form.

#### A. Scenario i): Mixed states with fixed purity

Adopting the Bloch sphere representation we express the template states  $\rho_1$  and  $\rho_2$  in terms of their associated Bloch vectors  $\mathbf{r}_1$  and  $\mathbf{r}_2$  via the mapping

$$\rho_1 = \frac{\mathbf{1} + \mathbf{r}_1 \cdot \boldsymbol{\sigma}}{2}, \quad \rho_2 = \frac{\mathbf{1} + \mathbf{r}_2 \cdot \boldsymbol{\sigma}}{2}, \quad (23)$$

with  $\boldsymbol{\sigma} = (\sigma_x, \sigma_y, \sigma_z)$  being the Pauli vector. Assuming then the purity of these density matrices to be assigned, we keep the modulus  $r_1 \equiv |\mathbf{r}_1|$  and  $r_2 \equiv |\mathbf{r}_2|$  constant and use  $d\mu(\rho_1, \rho_2)$  to average over all possible orientations of  $\mathbf{r}_1$  and  $\mathbf{r}_2$  by setting it equal to

$$d\mu(\rho_1, \rho_2) = dU_1 dU_2, \quad (24)$$

with  $dU$  representing the Haar measure on the unitary transformations of  $SU(2)$ . Accordingly we rewrite Eq. (4) as

$$\alpha^{(n)} = \int dU_1 \left( U_1 \rho_1 U_1^\dagger \right)^{\otimes n+1} \otimes \int dU_2 \left( U_2 \rho_2 U_2^\dagger \right)^{\otimes n}, \quad (25)$$

$$\beta^{(n)} = \int dU_1 \left( U_1 \rho_1 U_1^\dagger \right)^{\otimes n} \otimes \int dU_2 \left( U_2 \rho_2 U_2^\dagger \right)^{\otimes n+1}. \quad (26)$$

With this choice both  $\alpha^{(n)}$  and  $\beta^{(n)}$ , as well as their difference  $\Theta$ , become explicitly invariant under global rotations  $U^{\otimes 2n+1}$ . Therefore the eigenvectors of each one of these operators must be also eigenvectors of the total angular momentum of the total system  $AXB$ . Furthermore, we notice that  $\alpha^{(n)}$  and  $\beta^{(n)}$  are also invariant under separate rotations of partitions of the system, in particular  $AX/B$  for  $\alpha^{(n)}$  and  $A/XB$  for  $\beta^{(n)}$ . This allows us to cast the first of the above equations in the following form

$$\alpha^{(n)} = \oplus_{s',t} f_{s'}^{(n+1)}(r_1) f_t^{(n)}(r_2) \mathbf{1}_{AX}^{(s')} \otimes \mathbf{1}_B^{(t)}, \quad (27)$$

where  $\mathbf{1}_Q^{(j)}$  indicates the projector on all the irreducible representations in the system  $Q$  with Casimir  $j(j+1)$  (i.e. the space  $\mathcal{H}_B^{(t)}$  of (12) for  $\mathbf{1}_B^{(t)}$ , and the subspaces of  $\mathcal{H}_A^{(s)} \otimes \mathcal{H}_X$  spanned by vectors with Casimir  $s' = s \pm 1/2$  for  $\mathbf{1}_{AX}^{(s')}$ ), and where the non-negative function

$$f_j^{(n)}(r) = \frac{1}{2j+1} \left( \frac{1-r^2}{4} \right)^{\frac{n}{2}-j} \times \frac{\left( \frac{1+r}{2} \right)^{2j+1} - \left( \frac{1-r}{2} \right)^{2j+1}}{r}, \quad (28)$$

encodes the dependence upon  $r_1$  and  $r_2$  (see Appendix VI-A for details). Adopting the basis (13) we can then use Eq. (27) to decompose  $\alpha^{(n)}$  as a direct sum of independent contributions acting on the subspaces  $\mathcal{H}_{AXB_k}^{(s,t)}$ , i.e.

$$\alpha^{(n)} = \oplus_{s,t} \oplus_{i,k} \left( \oplus_{q,m} \alpha^{(n)}|_{i,k}^{s,t,q,m} \right), \quad (29)$$

where, for each  $s, t, i$  and  $k$  we exploited the fact that each term further decompose into a direct sum of either  $1 \times 1$  or  $2 \times 2$  blocks of the form

$$\begin{aligned} \alpha^{(n)}|_{i,k}^{s,t,q,m} &= f_{s+1/2}^{(n+1)}(r_1) f_t^{(n)}(r_2) \\ &|s+1/2, t; q, m\rangle_{i,k} \langle s+1/2, t; q, m| \\ &+ f_{s-1/2}^{(n+1)}(r_1) f_t^{(n)}(r_2) \\ &|s-1/2, t; q, m\rangle_{i,k} \langle s-1/2, t; q, m|, \end{aligned} \quad (30)$$

where as already mentioned it is implicit assumed that the vectors  $|s \pm 1/2, t; q, m\rangle_{i,k}$  nullify whenever the parameters  $s, t$  and  $q$  do not fit the angular



momentum selection rules. In a similar fashion we have that

$$\beta^{(n)} = \oplus_{s,t'} f_s^{(n)}(r_1) f_{t'}^{(n+1)}(r_2) \mathbf{1}_A^{(s)} \otimes \mathbf{1}_{XB}^{(t')}, \quad (31)$$

where now  $\mathbf{1}_A^{(s)}$  project on  $\mathcal{H}_A^{(s)}$  of (12) and  $\mathbf{1}_{XB}^{(t')}$  on the subspaces of  $\mathcal{H}_X \otimes \mathcal{H}_B^{(t')}$  spanned by vectors with Casimir  $t' = t \pm 1/2$ . Again this yields the following decomposition

$$\beta^{(n)} = \oplus_{s,t} \oplus_{i,k} \left( \oplus_{q,m} \beta^{(n)}|_{i,k}^{s,t,q,m} \right), \quad (32)$$

where now

$$\begin{aligned} \beta^{(n)}|_{i,k}^{s,t,q,m} &= f_s^{(n)}(r_1) f_{t+1/2}^{(n+1)}(r_2) \\ &|s, t+1/2; q, m\rangle_{i,k} \langle s, t+1/2; q, m| \\ &+ f_s^{(n)}(r_1) f_{t-1/2}^{(n+1)}(r_2) \\ &|s, t-1/2; q, m\rangle_{i,k} \langle s, t-1/2; q, m|. \end{aligned} \quad (33)$$

In this expression the elements

$$\{|s, t' = t \pm 1/2; q, m\rangle_{i,k}\}_{q,m}, \quad (34)$$

are obtained by coupling the qubit  $\mathcal{H}_X$  with those of  $\mathcal{H}_{B,k}^{(t)}$  and, as usual, we assume they nullify whenever  $s, t$  and  $q$  do not fulfil the necessary selection rules. These vectors form a new basis for  $\mathcal{H}_{A_i X B_k}^{(s,t)}$  connected with the one presented in Eq. (13) via the following four amplitude probabilities

$$\begin{aligned} C_{++}^{(s,t,q)} &\equiv {}_{i,k} \langle s + \frac{1}{2}, t; q, m | s, t + \frac{1}{2}; q, m \rangle_{i,k} \\ C_{+-}^{(s,t,q)} &\equiv {}_{i,k} \langle s + \frac{1}{2}, t; q, m | s, t - \frac{1}{2}; q, m \rangle_{i,k} \\ C_{-+}^{(s,t,q)} &\equiv {}_{i,k} \langle s - \frac{1}{2}, t; q, m | s, t + \frac{1}{2}; q, m \rangle_{i,k} \\ C_{--}^{(s,t,q)} &\equiv {}_{i,k} \langle s - \frac{1}{2}, t; q, m | s, t - \frac{1}{2}; q, m \rangle_{i,k} \end{aligned} \quad (35)$$

which admit closed analytical expressions in terms of the Wigner  $6j$  coefficients, i.e.

$$\begin{aligned} C_{\pm\pm}^{(s,t,q)} &= (-1)^{\pm\frac{1}{2}\pm\frac{1}{2}} \\ &\times \sqrt{(2s \pm 1 + 1)(2t \pm 1 + 1)} \\ &\times \begin{Bmatrix} t \pm \frac{1}{2} & t & \frac{1}{2} \\ s \pm \frac{1}{2} & s & q \end{Bmatrix}. \end{aligned} \quad (36)$$

From Eqs. (29) and (32) it now follows that a similar decomposition holds also for  $\Theta$ ,

$$\Theta = \oplus_{s,t} \oplus_{i,k} \left( \oplus_{q,m} \Theta|_{i,k}^{s,t,q,m} \right), \quad (37)$$

where for assigned  $s, t, i$  and  $k$ ,  $\Theta|_{i,k}^{s,t,q,m}$  are the following  $1 \times 1$  or  $2 \times 2$  matrices

$$\Theta|_{i,k}^{s,t,q,m} = \alpha^{(n)}|_{i,k}^{s,t,q,m} - \beta^{(n)}|_{i,k}^{s,t,q,m}. \quad (38)$$

Invoking the convention established when introducing Eq. (13) we notices that  $1 \times 1$  blocks occur explicitly in the *a*), *b*) and *c*) scenarios detailed in the introductory part of the section, yielding the eigenvalues

$$\lambda_{s,t,q}^{(n)} = f_s^{(n)}(r_1) f_t^{(n)}(r_2) \Lambda_{s,t,q}^{(n)}, \quad (39)$$

with

$$\Lambda_{s,t,q}^{(n)} = \begin{cases} R_{s,+}^{(n)}(r_1) - R_{t,+}^{(n)}(r_2) & \text{case a),} \\ R_{s,+}^{(n)}(r_1) - R_{t,-}^{(n)}(r_2) & \text{case b),} \\ R_{s,-}^{(n)}(r_1) - R_{t,+}^{(n)}(r_2) & \text{case c),} \end{cases} \quad (40)$$

where we introduced the functions

$$R_{j,\pm}^{(n)}(r) \equiv \frac{f_{j\pm 1/2}^{(n+1)}(r)}{f_j^{(n)}(r)}. \quad (41)$$

For  $s, t$ , and  $q$  belonging to case *d*) instead, (38) is a  $2 \times 2$  matrix of the form (18) with eigenvalues as in (22) with the following identifications

$$\begin{aligned} \Theta_{++}^{(s,t,q)} &= f_s^{(n)}(r_1) f_t^{(n)}(r_2) \left[ R_{s,+}^{(n)}(r_1) \right. \\ &\quad \left. - R_{t,+}^{(n)}(r_2) (C_{++}^{(s,t,q)})^2 - R_{t,-}^{(n)}(r_2) (C_{+-}^{(s,t,q)})^2 \right], \\ \Theta_{--}^{(s,t,q)} &= f_s^{(n)}(r_1) f_t^{(n)}(r_2) \left[ R_{s,-}^{(n)}(r_1) \right. \\ &\quad \left. - R_{t,+}^{(n)}(r_2) (C_{-+}^{(s,t,q)})^2 - R_{t,-}^{(n)}(r_2) (C_{--}^{(s,t,q)})^2 \right], \end{aligned}$$

and

$$\begin{aligned} \Theta_{+-}^{(s,t,q)} &= -f_s^{(n)}(r_1) f_t^{(n)}(r_2) \\ &\left[ R_{t,+}^{(n)}(r_2) C_{++}^{(s,t,q)} C_{-+}^{(s,t,q)} + R_{t,-}^{(n)}(r_2) C_{+-}^{(s,t,q)} C_{--}^{(s,t,q)} \right], \end{aligned}$$

where we used the coefficients (35) to express the elements of  $\beta^{(n)}|_{i,k}^{s,t,q,m}$  into the basis (13). The corresponding eigenvalues can also be expressed as in the rescaled form (39) with

$$\Lambda_{s,t,q}^{(n)}(\pm) = a_{s,t}^{(n)} \pm b_{s,t,q}^{(n)}, \quad (42)$$

the functions  $a_{s,t}^{s,t}$  and  $b_q^{s,t}$  being defined as

$$a_{s,t}^{(n)} \equiv \frac{R_{s,+}^{(n)}(r_1) + R_{s,-}^{(n)}(r_1) - R_{t,+}^{(n)}(r_2) - R_{t,-}^{(n)}(r_2)}{2}, \quad (43)$$

$$b_{s,t,q}^{(n)} \equiv \frac{\sqrt{[G_s(r_1) - G_t(r_2)]^2 - 4G_s(r_1)G_t(r_2)(C_{++}^{(s,t,q)})^2}}{2}, \quad (44)$$

where for ease of notation we introduced

$$G_j(r) \equiv f_{j+1/2}^{(n+1)}(r) - f_{j-1/2}^{(n+1)}(r). \quad (45)$$

For future reference we observe that from Eq. (36) the following inequality can be determined

$$b_{s,t,q}^{(n)} \geq b_{s,t,q=s+t-1/2}^{(n)}, \quad (46)$$

which in turn can be used to establish useful bounds for the eigenvalues (42), i.e.

$$\Lambda_{s,t,q}^{(n)}(+) \geq \Lambda_{s,t,q=s+t-1/2}^{(n)}(+), \quad (47)$$

$$\Lambda_{s,t,q}^{(n)}(-) \leq \Lambda_{s,t,q=s+t-1/2}^{(n)}(-). \quad (48)$$

Replacing all this into Eq. (10) we can finally write

$$P_{err,min}^{(n)} = \frac{1}{2} - \frac{1}{2} \sum_{s,t,q,\ell}^+ f_s^{(n)}(r_1) f_t^{(n)}(r_2) M_{s,t,q}^{(n)} \Lambda_{s,t,q}^{(n)}(\ell), \quad (49)$$

with  $M_{s,t,q}^{(n)}$  being the multiplicity factor defined in Eq. (16), the index  $\ell$  assuming the values  $\pm$  for the case  $d$ ), and where the subscript  $+$  indicates that only the positive values of  $\Lambda_{s,t,q}^{(n)}(\ell)$  are allowed into the sum. In order to get an asymptotic expansion of Eq. (49) we now notice that for large  $n$  the following expansion holds,

$$f_s^{(n)}(r) \#(s, n) \approx \frac{1+r}{r} \frac{1}{1 + \frac{n}{2} + s} \times B(n, \frac{1+r}{2}, n/2 + s) \quad (50)$$

where  $B(n, \frac{1+r}{2}, n/2 + s)$  is a binomial distribution for the variable  $n/2 + s$ , and the neglected terms give an exponentially suppressed contribution as  $n$  goes to infinity. The mean of  $\frac{s}{n}$  is  $\frac{r}{2}$  and the variance is  $\frac{1-r^2}{4n}$ , the next moments give contribution  $O(n^{-2})$ . The sum on  $s$  goes to zero or  $1/2$  to  $n/2$ , therefore if  $r$  is sufficiently greater than 0 we are neglecting in the sum a region where the binomial distribution is small and the total contribution of the region to the sum is exponentially suppressed. The second useful observation is that the eigenvalues and the term outside the binomial in (50), expanded

in the variables  $\frac{s}{n}$  and  $\frac{t}{n}$  around their means, show series coefficients that do not increase in powers of  $n$  as one goes to higher terms. Therefore to get the leading and next to leading term one needs the expansion only at second order in these variables.

The expansion in  $\frac{s}{n}, \frac{t}{n}$  around their means let us also determine the sign of the eigenvalues in the relevant region for the sum. In particular for the four cases analyzed so far we have:

$$\begin{aligned} a) \quad & \Lambda_{s,t,q=s+t+1/2}^{(n)} = \frac{r_1-r_2}{2} + O\left(\frac{s}{n} - \frac{r_1}{2}, \frac{t}{n} - \frac{r_2}{2}\right), \\ b) \quad & \Lambda_{s,t,q=t-s-1/2}^{(n)} = \frac{r_1+r_2}{2} + O\left(\frac{s}{n} - \frac{r_1}{2}, \frac{t}{n} - \frac{r_2}{2}\right), \\ c) \quad & \Lambda_{s,t,q=s-t-1/2}^{(n)} = -\frac{r_1+r_2}{2} + O\left(\frac{s}{n} - \frac{r_1}{2}, \frac{t}{n} - \frac{r_2}{2}\right), \\ d) \quad & \Lambda_{s,t,q}^{(n)}(+) \geq \frac{\sqrt{(r_1-r_2)^2}}{2} + O\left(\frac{s}{n} - \frac{r_1}{2}, \frac{t}{n} - \frac{r_2}{2}\right), \\ & \Lambda_{s,t,q}^{(n)}(-) \leq -\frac{\sqrt{(r_1-r_2)^2}}{2} + O\left(\frac{s}{n} - \frac{r_1}{2}, \frac{t}{n} - \frac{r_2}{2}\right), \end{aligned}$$

where in deriving the last two terms we used (47) and (48). The above expressions allows us to identify the positive terms which, in the limit of large  $n$ , contribute to the sum (49): for instance taking  $r_1 > r_2$  we noticed that the positive eigenvalues are those associated with case  $a$ ) and the first of case  $d$ ), while the case  $b$ ), which is also positive, can be ignored because  $t > s$  is not in the relevant region of the sum on  $s, t$ . With this information, the sum on  $q$  can now be performed at the relevant order with the first order of the Euler-MacLaurin expansion. The final result, which takes into account also the case  $r_1 < r_2$ , is

$$\begin{aligned} P_{err,min}^{(n \gg 1)} & \simeq \frac{1}{2} - \frac{1}{24} \frac{(r_1 + r_2)^3 - |r_1 - r_2|^3}{r_1 r_2} \\ & + \frac{5}{24 n} \frac{(r_1 + r_2)^3 + |r_1 - r_2|^3}{r_1^2 r_2^2} \\ & + \frac{1}{24 n} \frac{(r_1 + r_2)^5 - |r_1 - r_2|^5}{r_1^3 r_2^3}. \end{aligned} \quad (51)$$

which for  $r_1 = r_2$  reproduce correctly the result of [19], and which in agreement with (9) exhibits a leading order that corresponds to the average of

the Helstrom probabilities, i.e.

$$\begin{aligned}\bar{P}_H &= \frac{1}{2} - \frac{1}{4} \int \sin \theta d\theta \frac{\sqrt{(r_1 - r_2 \cos \theta)^2 + r_2^2 \sin^2 \theta}}{2} \\ &= \frac{1}{2} - \frac{1}{24} \frac{(r_1 + r_2)^3 - |r_1 - r_2|^3}{r_1 r_2}.\end{aligned}\quad (52)$$

### B. Scenario ii): Mixed states with hard sphere prior

In the scenario ii) we are interested in considering the case where  $\rho_1$  and  $\rho_2$  are arbitrary (possibly) mixed density matrices. This corresponds to replace (24) with

$$d\mu(\rho_1, \rho_2) = dU_1 d\mu(r_1) dU_2 d\mu(r_2), \quad (53)$$

where again  $dU$  represents the Haar measure of  $SU(2)$  while  $d\mu(r)$  is a measure that gauges our ignorance about the purity of the template states, i.e. the length of their associated Bloch vectors. Accordingly the only difference with the previous paragraph is that now, in the expression of  $\alpha^{(n)}$  and  $\beta^{(n)}$  given in Eqs. (29) and (32) we have now to replace the functions  $f_j^{(n)}(r)$  with their averaged values, i.e.

$$f_j^{(n)}(r) \rightarrow f_j^{(n)} \equiv \int d\mu(r) f_j^{(n)}(r). \quad (54)$$

As a choice for  $d\mu(r)$  we take the hard sphere prior measure, i.e.

$$d\mu(r) = \frac{3}{4\pi} r^2 dr \quad (55)$$

which yields

$$f_j^{(n)} = 2 \frac{(\frac{n}{2} - s)! (1 + \frac{n}{2} + s)!}{(n+3)!}. \quad (56)$$

The associated eigenvalues of  $\Theta$  can then be expressed as in (39) with the rescaled quantities  $\Lambda_q^{s,t}$  obtained as before with the terms  $R_{j,\pm}^{(n)}(r)$  being replaced by

$$\begin{aligned}R_{s,+}^{(n)} &\equiv \frac{f_{s+1/2}^{(n+1)}}{f_s^{(n)}} = \frac{2 + \frac{n}{2} + s}{n+4}, \\ R_{s,-}^{(n)} &\equiv \frac{f_{s-1/2}^{(n+1)}}{f_s^{(n)}} = \frac{1 + \frac{n}{2} - s}{n+4}.\end{aligned}\quad (57)$$

As a result, for the cases *a)*, *b)*, *c)*, and *d)*, we get the following solutions,

$$\begin{aligned}a) \quad \Lambda_{s,t,s+t+\frac{1}{2}}^{(n)} &= \frac{s-t}{n+4}, \\ b) \quad \Lambda_{s,t,t-s-\frac{1}{2}}^{(n)} &= \frac{1+s+t}{n+4}, \\ c) \quad \Lambda_{s,t,s-t-\frac{1}{2}}^{(n)} &= -\frac{1+s+t}{n+4}, \\ d) \quad \Lambda_{s,t,q}^{(n)}(\pm) &= \pm \frac{\sqrt{3-4q(1+q)+8s(1+s)+8t(1+t)}}{2(n+4)},\end{aligned}$$

which shows that only terms entering in the expression (49) for  $P_{err,min}^{(n)}$  are those of *a)* with  $s > t$ , those of *b)*, and the  $\Lambda_{s,t,q}^{(n)}(+)$  term of *d)*. Accordingly we can write

$$P_{err,min}^{(n)} = \frac{1 - S^{(n)}}{2}, \quad (58)$$

with

$$\begin{aligned}S^{(n)} &= \sum_{s>t} f_s^{(n)} f_t^{(n)} M_{s,t,s+t+\frac{1}{2}}^{(n)} \Lambda_{s,t,s+t+\frac{1}{2}}^{(n)} \\ &\quad + \sum_{t>s} f_s^{(n)} f_t^{(n)} M_{s,t,t-s-\frac{1}{2}}^{(n)} \Lambda_{s,t,t-s-\frac{1}{2}}^{(n)} \\ &\quad + \sum_{s,t} f_s^{(n)} f_t^{(n)} \sum_{q=|s-t|+\frac{1}{2}}^{s+t-\frac{1}{2}} M_{s,t,q}^{(n)} \Lambda_{s,t,q}^{(n)}(+),\end{aligned}\quad (59)$$

with  $M_{s,t,q}^{(n)}$  the multiplicity factors of defined in Eq. (16) which allow for a simplification of the resulting formula thanks to the identity

$$f_s^{(n)} f_t^{(n)} M_{s,t,q}^{(n)} = \frac{4(2s+1)(2t+1)(2q+1)}{(n+1)^2(n+2)^2}. \quad (60)$$

To get to the final result at order  $O(\frac{1}{n})$  one can still exploit the Euler McLaurin formula for each of the three sums, but a careful inspection shows that the next-to-leading term is also needed:

$$\sum_{i=a}^b f(i) \approx \int_a^b f(x) dx + \frac{f(a) + f(b)}{2} \quad (61)$$

After a long calculation the result is

$$P_{err,min}^{(n \gg 1)} \simeq \frac{33}{70} + \frac{2}{35n}, \quad (62)$$

which in  $n \rightarrow \infty$  agrees with the average Helstrom probability  $\bar{P}_H = 33/70$  that in the present case



can be obtained by integrating (52) with respect to  $r_1$  and  $r_2$  with the corresponding hard sphere measures.

### C. Scenario iii): Pure states at fixed overlap

We now consider the case where the templates states  $\rho_1 = |\psi_1\rangle\langle\psi_1|$  and  $\rho_2 = |\psi_2\rangle\langle\psi_2|$  are pure and characterized by a mutual overlap which is known a priori. No information about the absolute orientation of the couple is instead assumed. As anticipated in the introductory section this model appears to be well suited to characterize a scenario where for instance the machine is asked to discriminate between two possible configurations on the basis of templates generated by an external party which does not share a common reference frame with the machine itself. Without loss of generality we can model this problem by setting

$$|\psi_1\rangle = |\uparrow\rangle, \quad |\psi_2\rangle = U_0|\uparrow\rangle, \quad (63)$$

with a fixed unitary  $U_0$ , and then average on the action of a global unitary transformation  $U$  on  $\rho_1, \rho_2$ . With this choice the states (4) become

$$\begin{aligned} \alpha^{(n)} &= \int dU (U|\uparrow\rangle\langle\uparrow|U^\dagger)^{\otimes n+1} \\ &\quad \otimes (UU_0|\uparrow\rangle\langle\uparrow|U_0^\dagger U^\dagger)^{\otimes n} \end{aligned} \quad (64)$$

$$\begin{aligned} \beta^{(n)} &= \int dU (U|\uparrow\rangle\langle\uparrow|U^\dagger)^{\otimes n} \\ &\quad \otimes (UU_0|\uparrow\rangle\langle\uparrow|U_0^\dagger U^\dagger)^{\otimes n+1} \end{aligned} \quad (65)$$

where once more  $dU$  is the Haar measure of  $SU(2)$ . As in the cases analyzed before  $\alpha^{(n)}$  and  $\beta^{(n)}$ , as well as their difference  $\Theta^{(n)}$  are invariant under global rotations. Furthermore, since on both the  $AX$  and the  $B$  partition  $\alpha^{(n)}$  is described by pure vectors which are completely symmetric under permutations, the only elements of the basis (13) on which it can have support are those with maximum values of  $s'$  and  $t$ , i.e.  $s' = (n+1)/2$ ,  $t = n/2$ . As explicitly derived in (VI-B), these states are also eigenstates for  $\alpha^{(n)}$ , i.e.

$$\alpha^{(n)}|\frac{n+1}{2}, \frac{n}{2}; q, m\rangle = \Phi^{(n)}(q, U_0)|\frac{n+1}{2}, \frac{n}{2}; q, m\rangle, \quad (66)$$

with eigenvalues  $\Phi^{(n)}(q, U_0)$  given by

$$\begin{aligned} \Phi^{(n)}(q, U_0) &= \sum_{h=-n/2}^{n/2} D_{h, \frac{n+1}{2}}^{\frac{n+1}{2}}(U_0) D_{\frac{n+1}{2}, h}^{\frac{n+1}{2}}(U_0^\dagger) \\ &\quad \times \frac{1}{2q+1} C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h} C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h}, \end{aligned} \quad (67)$$

where the symbol  $D_{mm'}^j(U)$  represent the matrix elements of the irreducible representations of  $U \in SU(2)$  with Casimir  $j(j+1)$ , and  $C_{j,m,j',m'}^{q,l}$  being the Clebsch-Gordan coefficients. (Notice that in the above analysis we dropped the multiplicity labels  $i$  and  $k$  in writing the elements of the basis (13) because for the  $s' = (n+1)/2$ ,  $t = n/2$  no degeneracy of the representation is present,  $\#(n/2, n) = 1$ ). Analogous properties applies for  $\beta^{(n)}$  when expressed into the basis (34). Therefore as in the previous cases  $\Theta$  can be expressed as a direct sum of  $1 \times 1$  and  $2 \times 2$  block matrices. In the present case, however due to the special restriction on  $s$  and  $t$  instead of the four possible cases observed in the previous section, only  $a)$  and  $d)$  may occur. It turns out that for the case  $a)$  the associated eigenvalues is always null. For  $d)$  instead we have

$$\begin{aligned} \lambda_{s=n/2, t=n/2, q}^{(n)}(\pm) &= \pm \Phi^{(n)}(q, U_0) \\ &\quad \times |C_{+-}^{(s=n/2, t=n/2, q)}|, \end{aligned} \quad (68)$$

and the eigenvectors are the same that we obtain for  $r_1 = r_2 = 1$  in the case of completely random orientations: for pure states, the optimal POVM in the fixed overlap case is the same. Replacing all this into Eq. (10) we can finally write

$$P_{err, min}^{(n)} = \frac{1}{2} - \frac{1}{2} \sum_q (2q+1) \lambda_{s=n/2, t=n/2, q}^{(n)}(+), \quad (69)$$

where we used the fact that  $M_{s=n/2, t=n/2, q}^{(n)} = 2q+1$  and that only the  $+$  elements of the couples (68) are positive. To proceed further, without loss of generality, we write  $U_0 = \exp(-i\sigma_y(\pi - \theta)/2)$  obtaining

$$\begin{aligned} D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, h}^{\frac{n}{2}}(U_0^\dagger) &= \frac{n!}{(\frac{n}{2}+h)!(\frac{n}{2}-h)!} \\ &\quad \times (\cos^2(\frac{\pi-\theta}{2}))^{\frac{n}{2}+h} (\sin^2(\frac{\pi-\theta}{2}))^{\frac{n}{2}-h}, \end{aligned} \quad (70)$$

which is a binomial distribution in the variable  $\frac{n}{2} + h \in \{0, n\}$ . We also notice that

$$\left(C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, h}^q\right)^2 = \frac{2(\frac{n}{2}-h)!(n+1)!}{(\frac{n}{2}+h)!} \times \frac{(\frac{n}{2}+h+q+\frac{1}{2})!}{(q-\frac{1}{2}-\frac{n}{2}-h)!(n-q+\frac{1}{2})!(n+q+\frac{3}{2})!}, \quad (71)$$

is also a probability distribution in the variable  $q \in \{\frac{n}{2} + h, n + \frac{1}{2}\}$ . Then the terms entering in the sum of Eq. (69) rewrite explicitly as

$$(2q+1)\lambda_{s=n/2, t=n/2, q}^{(n)}(+) = \sum_h \frac{n!}{(\frac{n}{2}+h)!(\frac{n}{2}-h)!} \times \left(\cos^2\left(\frac{\pi-\theta}{2}\right)\right)^{\frac{n}{2}+h} \left(\sin^2\left(\frac{\pi-\theta}{2}\right)\right)^{\frac{n}{2}-h} \times \frac{2(\frac{n}{2}-h)!(\frac{n}{2}+h+q+\frac{1}{2})!(n+1)!}{(\frac{n}{2}-h+q-\frac{1}{2})!(\frac{n}{2}+h)!(n-q+\frac{1}{2})!(n+q+\frac{3}{2})!} \times \frac{1}{2} \sqrt{\frac{2(3/2+q+n)(1/2-q+n)}{(n/2+1/2)(n+1)}}.$$

As usual we focus on the limit of large  $n \gg 1$  for  $P_{err, min}^{(n)}$ . In this case we notice that in order to get up to the order  $O(\frac{1}{n^2})$  for the resulting expression, one can expand  $|C_{s=n/2, t=n/2, q}^{(n)}|$  around the mean of the  $q$  distribution and consider contributions up to the fourth central moment (see Appendix VI-C), expand the result around the mean of the  $h$  distribution and calculate the contributions up to the relevant moment (not more than the fourth). The result is

$$P_{err, min}^{(n \gg 1)} \simeq \frac{1}{2} \left(1 - |\cos \frac{\theta}{2}|\right) + \frac{3+\cos \theta}{8\sqrt{2}\sqrt{1+\cos \theta}} \frac{1}{n} + \frac{1-60\cos \theta - 5\cos 2\theta}{128\sqrt{2}(1+\cos \theta)^{3/2}} \frac{1}{n^2}, \quad (72)$$

where, as expected, the first contribution corresponds to the corresponding averaged Helstrom probability  $\bar{P}_H$ . We also notice that for small deviations from orthogonality, one has

$$P_{err, min}^{(n \gg 1)} \simeq \frac{\theta^2}{16} + \frac{1}{4n} - \frac{1}{8n^2} \left(1 - \frac{\theta^2}{4}\right), \quad (73)$$

The expansion around coincident states is instead singular, but the formula is still valid when the states are not coincident and  $n(\pi - \theta) \gg 1$ . Since the optimal POVM is the same of the totally random pure state scenario, averaging over  $\theta$  before doing the asymptotic expansion gives the result of (51) when  $r_1 = r_2$ . Integrating at the end gives also

the same result up to first order, while the order  $n^{-2}$  is not integrable. This is not inconsistent: one can see that the averaged  $P_{err, min}^{(n)}$  displays a  $n^{-\frac{3}{2}}$  dependence which is not recoverable from this expansion (and also not exactly computable with the Euler-MacLaurin approximation), which at fixed  $n$  works only in the region  $n(\pi - \theta) \gg 1$ .

#### IV. IMPLEMENTATION OF THE OPTIMAL POVM

From the knowledge of the eigenvectors (21) one can reconstruct the optimal POVM. Since it is a projective measurement, it can be realized by a change of basis from the the eigenvectors to the computational basis, followed by a local measurement. The change of basis is the following:

$$\begin{aligned} |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{3}{2}, \frac{3}{2}}\rangle &\rightarrow |\uparrow\uparrow\uparrow\rangle \quad (C) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{3}{2}, \frac{1}{2}}\rangle &\rightarrow |\uparrow\uparrow\downarrow\rangle \quad (C) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}}^{(-)}\rangle &\rightarrow |\downarrow\uparrow\uparrow\rangle \quad (B) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}}^{(+)}\rangle &\rightarrow |\uparrow\downarrow\uparrow\rangle \quad (A) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, -\frac{1}{2}}^{(+)}\rangle &\rightarrow |\downarrow\downarrow\downarrow\rangle \quad (A) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, -\frac{1}{2}}^{(-)}\rangle &\rightarrow |\uparrow\downarrow\downarrow\rangle \quad (B) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{3}{2}, -\frac{1}{2}}\rangle &\rightarrow |\downarrow\downarrow\uparrow\rangle \quad (C) \\ |\psi_{\frac{1}{2}, \frac{1}{2}, \frac{3}{2}, -\frac{3}{2}}\rangle &\rightarrow |\downarrow\downarrow\downarrow\rangle \quad (C) \end{aligned} \quad (74)$$

where  $A$  ( $B$ ) means that the result of the measurement is interpreted as  $X = A$  ( $X = B$ ), while for  $C$  we "flip a coin" to decide. In the computational basis the unitary rotation reads

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & 0 & \frac{1}{\sqrt{3}} & 0 & 0 \\ 0 & \frac{1}{\sqrt{3}} & \frac{-3-\sqrt{3}}{6} & 0 & \frac{3-\sqrt{3}}{6} & 0 & 0 \\ 0 & 0 & 0 & \frac{-3-\sqrt{3}}{6} & 0 & \frac{1}{3+\sqrt{3}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{1}{\sqrt{3}} & \frac{3-\sqrt{3}}{6} & 0 & \frac{-3-\sqrt{3}}{6} & 0 & 0 \\ 0 & 0 & 0 & \frac{3-\sqrt{3}}{6} & 0 & \frac{1}{-3+\sqrt{3}} & \frac{1}{\sqrt{3}} \\ 0 & 0 & 0 & \frac{1}{\sqrt{3}} & 0 & \frac{1}{\sqrt{3}} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}, \quad (75)$$

and the probability of error as a function of  $\theta$  is

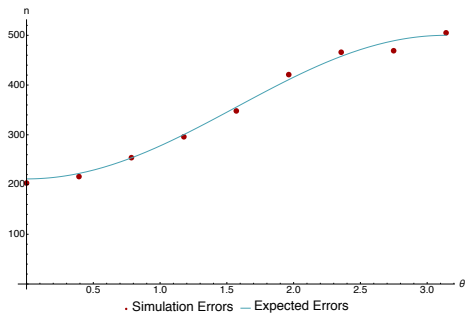


Fig. 1. Misclassification errors with 1000 repetitions for each angle, compared with the predicted minimum error function.

$$P_{err,min}^{(1)} = \frac{1}{2} - \frac{1 + \cos \theta}{4\sqrt{3}}. \quad (76)$$

These kind of operations are suitable for all programmable devices which are based on the circuit model of quantum computation, as for example the recent quantum chips developed by IBM [29]. By using the software development kit QISKit [30], we have determined a circuit that realizes the POVM for the  $n = 1$  case with input pure states and checked its performance with the IBM simulator. The circuit simulation and the theoretical prediction are shown in Figure IV.

We also report that we tried to remotely perform the experiment on the real physical chip but, unfortunately, the error propagation due to the large number of CNOT gates involved in our circuit was too strong to produce significant results. This fact underlines the importance of gate optimization and error correction for the proper operation of future quantum computers.

## V. CONCLUSIONS

In this manuscript we have discussed the performances of optimal universal learning quantum machines that aim at discriminating the states of a qubit starting from a collection of templates states in the hybrid, yet realistic scenario, where at least some global information on the training set is classically available. As a matter of fact, it is not hard to identify situations for which this kind

of approach could provide a realistic modelization. Indeed, while absolute information about quantum states is typically not accessible, some structural properties are more likely to be available. For instance this what happens in quantum communication [28] where the receiving party does not know the particular state is going to receive, but has classical knowledge on the code the sender is using. Our work could be extended along several directions, for example by assuming systems with a larger Hilbert space dimension or by allowing for more than just two template states.

## REFERENCES

- [1] S. Shalev-Shwartz and S. Ben-David, *Understanding Machine Learning: From Theory to Algorithms*. (Cambridge University Press, New York, NY, USA, 2014).
- [2] P. Wittek, *Quantum machine learning: what quantum computing means to data mining* (Academic Press, 2014).
- [3] M. Schuld, I. Sinayskiy, and F. Petruccione, *Contemporary Physics* **56.2**, 172-185 (2015).
- [4] V. Dunjko and H. J. Briegel, Eprint arXiv:1709.02779.
- [5] J. Biamonte, P. Wittek, N. Pancotti, P. Rebentrost, N. Wiebe and S. Lloyd, *Nature* **549**, 195-202 (2017).
- [6] C. Ciliberto, M. Herbster, A. D. Ialongo, M. Pontil, A. Rocchetto, S. Severini, L. Wossnig *Proc. R. Soc. A* **474** 20170551 (2018).
- [7] A. Monràs, G. Sentís, and P. Wittek, *Phys. Rev. Lett.*, **118** 190503 (2017).
- [8] M. Sasaki, A. Carlini, and R. Jozsa, *Phys. Rev. A*, **64** 022317 (2001).
- [9] M. Sasaki and A. Carlini, *Phys. Rev. A*, **66** 022303 (2002).
- [10] A. Hayashi, M. Horibe, and T. Hashimoto, *Phys. Rev. A*, **72** 052306 (2005).
- [11] Hayashi, M. Horibe, and T. Hashimoto, *Phys. Rev. A*, **73** 012328 (2006).
- [12] S. Gambs, Eprint arXiv:0809.0444.
- [13] C.W. Helstrom, *Quantum Detection and Estimation Theory* (Academic, New York, 1976).
- [14] A. S. Holevo, *J. Multivariate Anal.* **3**, 337 (1973).
- [15] H. P. Yuen, R. S. Kennedy and M. Lax, *IEEE Trans. Info. Theory* **21**, 125 (1975).
- [16] W. K. Wootters and W. H. Zurek, *Nature* **299**, 802 (1982).
- [17] J. A. Bergou and M. Hillery, *Phys. Rev. Lett.*, **94**, 160501 (2005).
- [18] M. Guță and W. Kotłowski, *New Journal of Physics*, **12** (12) 123032 (2010).
- [19] G. Sentís, E. Bagan, J. Calsamiglia, and R. Muñoz-Tapia, *Phys. Rev. A* **82** 042312 (2010).
- [20] G. Sentís, J. Calsamiglia, R. Muñoz-Tapia, *Scientific Reports*, **2** 708 (2012).
- [21] G. Sentís, E. Bagan, J. Calsamiglia and R. Muñoz-Tapia, *Phys. Rev. A*, **88** 052304 (2013).
- [22] G. Sentís, M. Guță, and G. Adesso, *EPJ Quantum Technology*, **2** (17) (2015).

- [23] A. Peres and W. K. Wootters, Phys. Rev. Lett., **66**, 1119 (1991).
- [24] S. Massar and S. Popescu, Phys. Rev. Lett., **74**, 1259 (1995).
- [25] N. Gisin and S. Popescu, Phys. Rev. Lett., **83**, 432 (1999).
- [26] S. Massar, Phys. Rev. A, **62**, 040101 (2000).
- [27] G. Chiribella, G. M. D'Ariano, P. Perinotti, and M. F. Sacchi, Phys. Rev. Lett. **93**, 180503 (2004)
- [28] A. S. Holevo, *Quantum Systems, Channels, Information* (de Gruyter Studies in Mathematical Physics, 2012).
- [29] IBM, *Quantum Experience*, URL: <https://www.research.ibm.com/ibm-q>.
- [30] IBM, *Quantum Information Software Kit (QISKit)*, URL: <https://www.qiskit.org>.

## VI. APPENDIX

In this appendix we present the explicit derivation of some important identities which are used in the main text. We recall that the Haar measure  $dU$  of  $SU(2)$  satisfies the identity

$$\int dU \mathcal{F}(LU) = \int dU \mathcal{F}(U) = \int dU \mathcal{F}(UR),$$

for all  $L, R \in SU(2)$  and for all functions  $\mathcal{F}$  acting on  $SU(2)$ , and that it induces a Hilbert product on  $L_2[U, dU]$  via the identification

$$(f, g) \equiv \int dU f^*(U)g(U).$$

Furthermore, indicating with  $D_{m, m'}^j(U)$  the matrix elements of the irreducible representations of  $U \in SU(2)$  with Casimir  $j(j+1)$ , we recall that via the Peter-Weyl theorem they fulfil the identities

$$\begin{aligned} \int dU \left( D_{m_1, m'_1}^{j_1}(U) \right)^* D_{m_2, m'_2}^{j_2}(U) \\ = \frac{1}{2j_1 + 1} \delta_{j_1 j'_1} \delta_{m_1 m_2} \delta_{m'_1 m'_2}. \end{aligned} \quad (77)$$

### A. Derivation of Eq. (27)

Let  $\rho$  a qubit density matrix characterized by Bloch vector of length  $r$  which, without loss of generality we shall assume to be oriented in the positive  $\hat{z}$  direction, i.e.  $\rho = \left(\frac{1+r}{2}\right) |\uparrow\rangle\langle\uparrow| + \left(\frac{1-r}{2}\right) |\downarrow\rangle\langle\downarrow|$  with  $|\uparrow\rangle, |\downarrow\rangle$  being the eigenvectors of  $\sigma_z$ . We notice that its  $n$ -th tensor power can be expressed as

$$\rho^{\otimes n} = \sum_{l=0}^n \left(\frac{1+r}{2}\right)^l \left(\frac{1-r}{2}\right)^{n-l} B_l^{(n)},$$

with

$$B_l^{(n)} \equiv \sum_{\pi} S_{\pi} (|\uparrow\rangle\langle\uparrow|^{\otimes l} \otimes |\downarrow\rangle\langle\downarrow|^{\otimes n-l}) S_{\pi}^{\dagger},$$

the sum being performed over the set of permutations operators  $S_{\pi}$  of  $n$  elements. By construction  $B_l^{(n)}$  is the projector on the eigenspace at fixed total angular momentum  $J_z$ , therefore it is diagonal in every basis of eigenvectors of  $J^2, J_z$ . In particular its support is given by the vectors  $|j, l - \frac{n}{2}\rangle_i$  in each representation with Casimir number  $J^2 = j(j+1)$  and  $l \in \{\frac{n}{2} - j, \dots, \frac{n}{2} + j\}$ , the index  $i$  labelling accounting for the multiplicity of the representation, i.e.

$$B_l^{(n)} = \oplus_{j \geq |l - \frac{n}{2}|} \oplus_i |j, l - \frac{n}{2}\rangle_i \langle j, l - \frac{n}{2}|. \quad (78)$$

Consider then the operator

$$\begin{aligned} \gamma^{(n)} &\equiv \int dU (U \rho U^{\dagger})^{\otimes n} \\ &= \sum_{l=0}^n \left(\frac{1+r}{2}\right)^l \left(\frac{1-r}{2}\right)^{n-l} P_l^{(n)}, \end{aligned} \quad (79)$$

with

$$P_l^{(n)} \equiv \int dU U^{\otimes n} B_l^{(n)} U^{\dagger \otimes n}. \quad (80)$$

Invoking the identity (77) we can conclude that

$$P_l^{(n)} = \oplus_{j \geq |l - \frac{n}{2}|} \frac{\mathbf{1}^{(j)}}{2j+1}, \quad (81)$$

where now  $\mathbf{1}^{(j)}$  is the projector on all irreducible representation with principal quantum number  $j$ . Accordingly we have

$$\begin{aligned} \gamma^{(n)} &= \oplus_j \sum_{l=\frac{n}{2}-j}^{\frac{n}{2}+j} \left(\frac{1+r}{2}\right)^l \left(\frac{1-r}{2}\right)^{n-l} \frac{\mathbf{1}^{(j)}}{2j+1} \\ &= \oplus_j f_j^{(n)}(r) \mathbf{1}^{(j)}, \end{aligned} \quad (82)$$

with  $f_j^{(n)}(r)$  as in (28). Equation (27) finally follows from by a direct application of (82) to the terms  $\int dU (U \rho_1 U^{\dagger})^{\otimes n+1}$  and  $\int dU (U \rho_2 U^{\dagger})^{\otimes n}$  that enter in the definition of the operator  $\alpha^{(n)}$  of Sec. III-A.

### B. Derivation of Eq. (67)

To derive (67) let us first expand  $|\frac{n+1}{2}, \frac{n}{2}; q, m\rangle$  into the angular momentum basis given by the tensor product states  $|\frac{n+1}{2}, m'\rangle \otimes |\frac{n}{2}, m-m'\rangle$  associated with the  $AX/B$  partition, i.e.

$$|\frac{n+1}{2}, \frac{n}{2}; q, m\rangle = \sum_{m'} C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m} \quad (83)$$

$$\times |\frac{n+1}{2}, m'\rangle \otimes |\frac{n}{2}, m-m'\rangle,$$

where  $C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m}$  are the corresponding Clebsch-Gordan coefficients. Then observing that in this basis the state  $|\uparrow\rangle^{\otimes n+1} \otimes |\uparrow\rangle^{\otimes n}$  corresponds to the element  $|\frac{n+1}{2}, \frac{n+1}{2}\rangle \otimes |\frac{n}{2}, \frac{n}{2}\rangle$ , we write the operator  $\alpha^{(n)}$  as

$$\alpha^{(n)} = \int dU U^{\otimes n+1} |\frac{n+1}{2}, \frac{n+1}{2}\rangle \langle \frac{n+1}{2}, \frac{n+1}{2}| U^{\dagger \otimes n+1}$$

$$\otimes (U U_0)^{\otimes n} |\frac{n}{2}, \frac{n}{2}\rangle \langle \frac{n}{2}, \frac{n}{2}| (U_0^\dagger U^\dagger)^{\otimes n},$$

and observe that

$$\alpha^{(n)} (|\frac{n+1}{2}, m'\rangle \otimes |\frac{n}{2}, m-m'\rangle) = \quad (84)$$

$$= \sum_{l, l'} \int dU D_{l, \frac{n+1}{2}}^{\frac{n+1}{2}}(U) D_{\frac{n+1}{2}, m'}^{\frac{n+1}{2}}(U^\dagger)$$

$$\times D_{l', \frac{n}{2}}^{\frac{n}{2}}(U U_0) D_{\frac{n}{2}, m-m'}^{\frac{n}{2}}(U_0^\dagger U^\dagger) |\frac{n+1}{2}, l\rangle \otimes |\frac{n}{2}, l'\rangle$$

$$= \sum_{h, k} D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, k}^{\frac{n}{2}}(U_0^\dagger)$$

$$\times \sum_{l, l'} \int dU D_{l, \frac{n+1}{2}}^{\frac{n+1}{2}}(U) D_{\frac{n+1}{2}, m'}^{\frac{n+1}{2}}(U^\dagger)$$

$$\times D_{l', h}^{\frac{n}{2}}(U) D_{k, m-m'}^{\frac{n}{2}}(U^\dagger) |\frac{n+1}{2}, l\rangle \otimes |\frac{n}{2}, l'\rangle,$$

where in the first identity the matrix elements

$$D_{m_1, m_2}^{\frac{n+1}{2}}(U) = \langle \frac{n+1}{2}, m_1 | U^{\otimes n+1} | \frac{n+1}{2}, m_2 \rangle,$$

and

$$D_{m_1, m_2}^{\frac{n}{2}}(U) = \langle \frac{n}{2}, m_1 | U^{\otimes n} | \frac{n}{2}, m_2 \rangle,$$

represent the action of the unitary  $U$  into the selected basis, while in the second we used the composition rules of  $SU(2)$  to factorize the contributions of  $U_0$  from the rest. This equation can be further simplified by exploiting once more the Clebsch-Gordan mapping (83) to merge together

$D_{l, \frac{n+1}{2}}^{\frac{n+1}{2}}(U)$  with  $D_{l', h}^{\frac{n}{2}}(U)$ , and  $D_{\frac{n+1}{2}, m'}^{\frac{n+1}{2}}(U^\dagger)$  with  $D_{k, m-m'}^{\frac{n}{2}}(U^\dagger)$ . As a result the previous expression becomes

$$\alpha^{(n)} (|\frac{n+1}{2}, m'\rangle \otimes |\frac{n}{2}, m-m'\rangle)$$

$$= \sum_{h, k} D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, k}^{\frac{n}{2}}(U_0^\dagger)$$

$$\times \sum_{l, l'} \int dU D_{l+ l', \frac{n+1}{2}+h}^q(U) D_{\frac{n+1}{2}+k, m}^q(U^\dagger)$$

$$\times C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m} C_{\frac{n+1}{2}, \frac{n}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h}$$

$$\times C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, k}^{q, \frac{n+1}{2}+k} C_{\frac{n+1}{2}, l, \frac{n}{2}, l'}^{q, l+ l'}$$

$$|\frac{n+1}{2}, l\rangle \otimes |\frac{n}{2}, l'\rangle$$

$$= \frac{1}{2q+1} \sum_{h, k} D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, k}^{\frac{n}{2}}(U_0^\dagger)$$

$$\times \sum_{l, l', q} C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m} C_{\frac{n+1}{2}, \frac{n}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h}$$

$$\times C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, k}^{q, \frac{n+1}{2}+h} C_{\frac{n+1}{2}, l, \frac{n}{2}, l'}^{q, m}$$

$$|\frac{n+1}{2}, l\rangle \otimes |\frac{n}{2}, l'\rangle,$$

where in the second identity we exploit the Peter-Weyl theorem, see Eq. (77), to evaluate the integral in  $U$ . Multiplying this by  $C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m}$  while summing over  $m'$ , the latter equation finally gives us

$$\alpha^{(n)} |\frac{n+1}{2}, \frac{n}{2}; q, m\rangle$$

$$= \frac{1}{2q+1} \sum_{h, k} D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, k}^{\frac{n}{2}}(U_0^\dagger)$$

$$\times \sum_{l, l', m', q'} C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q, m} C_{\frac{n+1}{2}, m', \frac{n}{2}, m-m'}^{q', m}$$

$$\times C_{\frac{n+1}{2}, \frac{n}{2}, \frac{n}{2}, h}^{q', \frac{n+1}{2}+h} C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, k}^{q', \frac{n+1}{2}+h}$$

$$\times C_{\frac{n+1}{2}, l, \frac{n}{2}, l'}^{q', m} |\frac{n+1}{2}, l\rangle \otimes |\frac{n}{2}, l'\rangle,$$

$$= \frac{1}{2q+1} \sum_h D_{h, \frac{n+1}{2}}^{\frac{n+1}{2}}(U_0) D_{\frac{n+1}{2}, h}^{\frac{n+1}{2}}(U_0^\dagger)$$

$$\times C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h} C_{\frac{n+1}{2}, \frac{n+1}{2}, \frac{n}{2}, h}^{q, \frac{n+1}{2}+h}$$

$$|\frac{n+1}{2}, \frac{n}{2}; q, m\rangle, \quad (85)$$

which coincides with (67).

### C. Central momenta of the distributions

Here we report the central momenta of the distributions used for computing (72).

Putting  $h = \frac{ns}{2}$  and  $\frac{1+r}{2} = \cos\left(\frac{\pi-\theta}{2}\right)$  we notice that the distribution

$$P_{\theta}^{(n)}(s) \equiv D_{h, \frac{n}{2}}^{\frac{n}{2}}(U_0) D_{\frac{n}{2}, h}^{\frac{n}{2}}(U_0^{\dagger}), \quad (86)$$

defined in Eq. (70) has momenta

$$\begin{aligned} \mu_1 &= E[s] = r, \\ \mu_2 &= E[(s - \mu_1)^2] = \frac{1 - r^2}{n}, \\ \mu_3 &= E[(s - \mu_1)^3] = 2r \frac{1 - r^2}{n}, \\ \mu_4 &= E[(s - \mu_1)^4] \\ &= \frac{(-1 + r^2)(2 - 6r^2 + 3n(-1 + r^2))}{n^3}. \end{aligned}$$

Instead, setting  $h = \frac{ns}{2}$ , we notice that the momenta of the distribution

$$\begin{aligned} P_h^{(n)}(q) &\equiv \frac{2\left(\frac{n}{2}-h\right)!(n+1)!}{\left(\frac{n}{2}+h\right)!} \\ &\times \frac{\left(\frac{n}{2}+h+q+\frac{1}{2}\right)!}{\left(q-\frac{1}{2}-\frac{n}{2}-h\right)!(n-q+\frac{1}{2})!(n+q+\frac{3}{2})!}, \end{aligned} \quad (87)$$

defined in (71), can be expressed in terms of Euler gamma functions as follows

$$\begin{aligned} \mu_1 &= E[q] \\ &= -\frac{1}{2} + \frac{\Gamma(1/2 + h + n/2)\Gamma(2 + n)}{\Gamma(1 + h + n/2)\Gamma(3/2 + n)} \\ &= \frac{n\sqrt{1+s}}{\sqrt{2}} - \frac{1}{2} \\ &+ \frac{11 + 5s}{8\sqrt{2}\sqrt{1+s}} + \frac{9 + 14s - 23s^2}{128\sqrt{2}n(1+s)^{3/2}} \\ &+ O\left(\frac{1}{n^2}\right), \end{aligned} \quad (88)$$

$$\begin{aligned} \mu_2 &= E[(q - \mu_1)^2] \\ &= \frac{1}{2}(1 + n)(2 + 2h + n) \\ &- \frac{\Gamma(3/2 + h + n/2)^2\Gamma(2 + n)^2}{\Gamma(1 + h + n/2)^2\Gamma(3/2 + n)^2} \\ &= \frac{1}{8}n(1 - s) + \frac{-1 + 2s - s^2}{64(1 + s)} + O\left(\frac{1}{n}\right) \end{aligned} \quad (89)$$

$$\begin{aligned} \mu_3 &= E[(q - \mu_1)^3] \\ &= \frac{-(8 + 2h(5 + 4n) + n(11 + 4n))\Gamma(3/2 + h + n/2)\Gamma(2 + n)}{\Gamma(1 + h + n/2)\Gamma(3/2 + n)} \\ &+ \frac{8\Gamma(3/2 + h + n/2)^3\Gamma(2 + n)^3}{4\Gamma(1 + h + n/2)^3\Gamma(3/2 + n)^3} \\ &= \frac{(-1 + s)^2 n}{32\sqrt{2}\sqrt{1 + s}} + O(1), \end{aligned} \quad (90)$$

$$\begin{aligned} \mu_4 &= E[(q - \mu_1)^4] \\ &= \frac{(1 + n)(4 + 10n + 4h^2n + 6n^2 + n^3 + 4h(1 + 3n + n^2))}{4} \\ &- \frac{3\Gamma(3/2 + h + n/2)^4\Gamma(2 + n)^4}{\Gamma(1 + h + n/2)^4\Gamma(3/2 + n)^4} \\ &+ \frac{(2 + 2n + n^2 + 2h(2 + n))\pi^2\Gamma(2 + 2h + n)^2\Gamma(3 + 2n)^2}{4^{3+2h+3n}\Gamma(1 + h + n/2)^4\Gamma(3/2 + n)^4} \\ &= \frac{3}{6}4(1 - 2s + s^2)n^2 + O(n). \end{aligned} \quad (91)$$