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Optimal arbitrarily accurate composite pulse sequences

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Implementing a single-qubit unitary is often hampered by imperfect control. Systematic amplitude errors ϵ , caused by incorrect duration or strength of a pulse, are an especially common problem. But a sequence of imperfect pulses can provide a better implementation of a desired operation, as compared to a single primitive pulse. We find optimal pulse sequences consisting of L primitive π or 2π rotations that suppress such errors to arbitrary order $O(\epsilon^n)$ on arbitrary initial states. Optimality is demonstrated by proving an L = O(n) lower bound and saturating it with L = 2n solutions. Closed-form solutions for arbitrary rotation angles are given for n = 1, 2, 3, 4. Perturbative solutions for any n are proven for small angles, while arbitrary angle solutions are obtained by analytic continuation up to n = 12. The derivation proceeds by a novel algebraic and nonrecursive approach, in which finding amplitude error correcting sequences can be reduced to solving polynomial equations.

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I. INTRODUCTION

Quantum computers are poised to solve a class of technologically relevant problems intractable on classical machines [1], but scalable implementations managing a useful number of qubits are directly impeded by two general classes of errors [2]. On one hand, unwanted system-bath interactions in open quantum systems lead to decoherence, and on the other, imperfect controls for addressing and manipulating qubit states result in cumulative errors that eventually render large computations useless.

Systematic amplitude errors, the consistent over- or underrotation of a single-qubit unitary operation by a small factor ϵ , are one common control fault. The discovery of a protocol for the complete and efficient suppression of these errors would greatly advance the field of quantum control, with applications as far ranging as implementing fault-tolerant quantum computation and improving nuclear magnetic resonance spectra acquisition. Due to the broad scope of systematic amplitude errors, this problem has been attacked repeatedly by a variety of methods with varying degrees of success [3–8]. A concept common to most approaches is the composite pulse sequence, in which some number of L carefully chosen erroneous primitive unitary operations, or pulses, are applied successively such that a target ideal rotation is approximated to some order n with an exponentially reduced error $O(\epsilon^{n+1})$.

In the realm of quantum computation, the criteria for useful pulse sequences are stringent: (1) For each order n, a procedure for constructing a pulse sequence correcting to that order is given. (2) This construction gives sequence lengths L that scale efficiently with n, that is, $L = O(n^k)$, with k as small as possible. (3) Sequences should be "fully compensating" or "class A" [4], meaning they operate successfully on arbitrary and unknown states (in contrast to "class B" sequences that operate successfully only on select initial states). (4) Although finite sets of universal quantum gates exist [1], ideally sequences should be capable of implementing arbitrary rotations so that quantum algorithms can be simplified conceptually and practically.

One finds that there are currently no sequences satisfying all four of these criteria and suppressing systematic amplitude errors. In the literature, SCROFULOUS [9], PB₁, BB₁ [10] satisfy

criteria (3) and (4) but offer corrections only up to order n=2. Unfortunately, generalizations of these to arbitrary n come with prohibitively long sequence lengths, so that criterion (2) ends up unsatisfied. Typically, a sequence correct to order n+1 is recursively constructed from those at order n, resulting in an inefficient sequence length $L=2^{O(n)}$ [5], although numerical studies suggest that efficient sequences with $L=O(n^{3.09})$ exist [5]. To date, other classes of systematic control errors [11–13] do not fare better.

There are some provable successes in efficient pulse sequences, though. However, to find them one must relax criterion (4), which requires arbitrary rotations. For example, if one restricts attention to correcting π rotations in the presence of amplitude errors, Jones proved the impressive result that sequences with $L = O(n^{1.47})$ [3,7] are possible. Uhrig efficiently implements the identity operator in the presence of dephasing errors with L = O(n) [14]. If we also relax the criterion (3) and settle for specialized class B sequences that take $|0\rangle$ to $|1\rangle$ (those we call inverting sequences), Vitanov has found efficient narrow-band sequences for amplitude errors also with L = O(n) [15]. Notably, both Uhrig's and Vitanov's results were achieved via algebraic, nonrecursive processes. In fact, as we show, a more generalized algebraic approach in the amplitude error case can reinstate the crucial criteria (3) and (4) while maintaining Vitanov's efficient length scaling.

Our main result is exactly such an algebraic generalization, a nonrecursive formalism for systematic amplitude errors. With this, we prove a lower bound of L=O(n) for class A sequences comprised of either primitive π or 2π rotations, then constructively saturate this bound to a constant factor with L=2n (plus a single initializing rotation). The improvement of these new sequences over prior state of the art is illustrated in Table I. We derive optimal closed-form solutions up to n=4 for arbitrary target angles and perturbative solutions for any n, valid for small target angles. We then analytically continue these perturbative solutions to arbitrary angles up to n=12. Since any random or uncorrected systematic errors in the primitive pulses accumulate linearly with sequence length, optimally short sequences such as ours minimize the effect of such errors.

TABLE I. Comparison of known pulse sequences operating on arbitrary initial states that suppress systematic amplitude errors to order n for arbitrary target rotation angles. Arbitrary accuracy generalizations are known or conjectured for all with the exception of SCROFULOUS. The sequences APn, PDn, and ToPn are presented in this work. Of interest is the subset of APn sequences labeled ToPn, for which arbitrary accuracy is provable perturbatively for small target angles.

Name	Length	notes
SCROFULOUS	3	$n = 1$, nonuniform θ_i [9]
Pn, Bn	$O(e^{n^2})$	Closed form [5]
SKn	$O(n^3)$	$n \leq 30$, Numerical [5]
		n > 30, Conjectured
APn (PDn)	2n	$n \leq 3(4)$, Closed form
		$n \leq 12$, Analytic continuation
		n > 12, Conjectured
ToPn	2n	Arbitrary n , perturbative

We define the problem statement for amplitude-error correcting pulse sequences mathematically in Sec. II, leading, in Sec. III, to a set of constraint equations that such pulse sequences must satisfy, which is then solved in Sec. IV by three approaches: analytical, perturbative, and numerical. The analytical method is interesting as it gives closed-form solutions for low order sequences in a systematic fashion. The perturbative method relies on the invertibility of the Jacobian of the constraints and is used for proving the existence of solutions for select target angles. The numerical method is the most straightforward and practical for higher orders, giving optimally short pulse sequences for correction orders up to n=12. Section V then presents several generalizations of our results, including discussions on narrow-band toggling, nonlinear amplitude errors, random errors, and simultaneous correction of off-resonance errors. Finally, we point out differences and similarities between our sequences and existing art in Sec. VI and conclude in Sec. VII.

II. PULSE SEQUENCES

A single-qubit rotation of target angle θ_T about the axis \vec{n} is the unitary $R_{\vec{n}}[\theta_T] = \exp[-i\theta_T(\vec{n}\cdot\vec{\sigma})/2]$, where $\vec{\sigma} = (\hat{X},\hat{Y},\hat{Z})$ is the vector of Pauli operators. Without affecting the asymptotic efficiency of our sequences, Euler angles allow us to choose $n_z = 0$, and consequently, we define $R_{\varphi}[\theta_T] = \exp(-i\theta_T\hat{\sigma}_{\varphi}/2)$ for $\hat{\sigma}_{\varphi} = \hat{X}\cos\varphi + \hat{Y}\sin\varphi$. However, we only have access to imperfect rotations $M_{\varphi}[\theta] = R_{\varphi}[(1+\epsilon)\theta]$ that overshoot a desired angle θ by $\epsilon\theta$, $|\epsilon| \ll 1$. With these primitive elements, we construct a pulse sequence \mathcal{S} consisting of L faulty pulses:

$$S = M_{\varphi_1}[\theta_1] M_{\varphi_2}[\theta_2] \dots M_{\varphi_L}[\theta_L]. \tag{1}$$

Denote by $\vec{\varphi}$ the vector of phase angles $(\varphi_1, \varphi_2, \ldots, \varphi_L)$, which are our free parameters. Leaving each amplitude θ_j as a free parameter (e.g., SCROFULOUS [9]) may help reduce sequence length, but we find that a fixed value $\theta_j = \theta_0$ leads to the most compelling results.

The goal is to implement a target rotation $R_{\varphi_0}[\theta_T]$ (or, without loss of generality, $R_0[\theta_T]$ by the replacement $\varphi_j \to$

 $\varphi_j - \varphi_0$), including the correct global phase, with a small error. The trace distance [1]

$$D(\hat{U}, \hat{V}) = \|\hat{U} - \hat{V}\| = \frac{1}{2} \text{Tr} \sqrt{(\hat{U} - \hat{V})^{\dagger} (\hat{U} - \hat{V})}$$
 (2)

is a natural metric for defining errors between two operators \hat{U} , \hat{V} [5]. We demand that the pulse sequence implements

$$S = R_0[-\epsilon \theta_T] + O(\epsilon^{n+1}), \tag{3}$$

so that the corrected rotation $U_T = \mathcal{S}M_0[\theta_T] = R_0[\theta_T] + O(\epsilon^{n+1})$ has trace distance with the same small leading error $D(U_T, R_0[\theta_T]) = O(\epsilon^{n+1})$. Thus constructed, U_T implements $R_0[\theta_T]$ over a very wide range of ϵ due to its first n derivatives vanishing and so has broadband characteristics [2].

For completeness, we mention other error quantifiers. First is the fidelity $F(\hat{U},\hat{V}) = \|\hat{U}\hat{V}^{\dagger}\|$, which is not truly a distance metric but can be easier to compute, and bounds $1 - F(\hat{U},\hat{V}) \leqslant D(\hat{U},\hat{V}) \leqslant \sqrt{1 - F(\hat{U},\hat{V})^2}$ [1]. The infidelity of U_T is then $1 - F(U_T,R_0\left[\theta_T\right]) = O(\epsilon^{2n+2})$, which is a commonly used quantifier [2,7]. Finally, for the specialized class B sequences called inverting sequences, the transition probability $|\langle 1|\hat{U}|0\rangle|^2$ is a viable quantity for comparison [15].

III. CONSTRAINT EQUATIONS

We now proceed to derive a set of equations, or constraints, on the phase angles $\vec{\varphi}$ that will yield broadband correction. We begin very generally in the first section by assuming just $\theta_j = \theta_0$ as mentioned before, but then we specialize in the subsequent two sections to the case $\theta_0 = 2\pi$ and the case of symmetric sequences, both of which greatly enhance tractability of the problem.

A. Equal amplitude base pulses

To begin, we obtain an algebraic expression for S by a direct expansion of a length L sequence. Defining $\theta'_0 = (1 + \epsilon)\theta_0/2$,

$$S = \prod_{j=1}^{L} M_{\varphi_j}[\theta_0] = \cos^L(\theta_0') \prod_{j=1}^{L} \left[1 - i \tan(\theta_0') \hat{\sigma}_{\varphi_j} \right]$$
$$= \sum_{j=0}^{L} A_L^j(\theta_0') \hat{\Phi}_L^j(\vec{\varphi}), \tag{4}$$

where indices in the matrix product ascend from left to right, $A_L^j(s) = (-i)^j \sin^j(s) \cos^{L-j}(s)$, and $\hat{\Phi}_L^j$ are noncommutative elementary symmetric functions generated by $\prod_{j=1}^L (1+t\hat{\sigma}_{\varphi_j}) = \sum_{j=0}^L t^j \hat{\Phi}_L^j$ [16]. The $\hat{\Phi}_L^j$ are hard to work with, so by applying the Pauli matrix identity,

$$\sigma_{\varphi_1}\sigma_{\varphi_2}\cdots\sigma_{\varphi_j} = \exp\left(i\hat{Z}\sum_{k=1}^{J}(-1)^k\varphi_k\right)\hat{X}^j,$$
 (5)

we obtain a more useful expression as functions of the phase angles φ_i :

$$\hat{\Phi}_{L}^{j}(\vec{\varphi}) = \left(\operatorname{Re} \left[\Phi_{L}^{j}(\vec{\varphi}) \right] I - i \operatorname{Im} \left[\Phi_{L}^{j}(\vec{\varphi}) \right] \hat{Z} \right) \hat{X}^{j}, \quad (6)$$

$$\Phi_L^j(\vec{\varphi}) = \sum_{1 \leqslant h_1 < h_2 < \dots < h_j \leqslant L} \exp\left(-i\sum_{k=1}^j (-1)^k \varphi_{h_k}\right). \quad (7)$$

By defining the terminal case $\Phi_L^0(\vec{\varphi}) \equiv 1$, the phase sums Φ_L^j are efficiently computable at numeric values of the phases by the recursion $\Phi_L^j = \Phi_{L-1}^j + \Phi_{L-1}^{j-1} e^{i(-1)^{j+1}\varphi_L}$ using dynamic programming (i.e., start from the terminal case and fill in the table Φ_L^j for all desired j and L).

Combining the expansion of S with Eq. (3) then imposes a set $\mathcal{B}_{n,L}$ of real constraints on $\vec{\varphi}$ to be satisfied by any order n, length L sequence. $\mathcal{B}_{n,L}$ is obtained by first matching coefficients of the trace orthogonal Pauli operators on either side of Eq. (3). We then obtain, in terms of normalized error $x = \frac{1}{2}\epsilon\theta_0$ and normalized target angle $\gamma = \theta_T/\theta_0$, the necessary and sufficient conditions

$$\sum_{j \in \begin{cases} \text{even} \\ \text{odd} \end{cases}} A_L^j(\theta_0/2 + x) \Phi_L^j(\vec{\varphi}) = \begin{cases} \cos(x\gamma) \\ i \sin(x\gamma) \end{cases} + O(x^{n+1}). \tag{8}$$

Second, the complex coefficients of x^0, x^1, \ldots, x^n are matched, giving 2(n+1) complex equations linear in the phase sums $\Phi_L^j(\vec{\varphi})$, or, letting $|\cdot|$ denote the size of a set, $|\mathcal{B}_{n,L}| = 4(n+1)$ real constraints.

However, these constraints $\mathcal{B}_{n,L}$ are intractable to direct solution, and a simplifying assumption is necessary. It should be reasonable to suspect that the small rotation $R_0\left[-\epsilon\theta_T\right]$ can be generated by small pure error terms $R_{\varphi}\left[\epsilon\theta_0\right]$. We will therefore set $\theta_0=2\pi$ [5]. Note that $\theta_0=\pi$ is also a tractable case but is related to the 2π -pulse case by phase toggling and so need not be considered separately. We give more detail on toggling in Sec. V.

B. Assuming base pulses of $\theta_0 = 2\pi$

We now enumerate several key results, due simply to imposing $\theta_0 = 2\pi$, that apply to all order n, length L, 2π -pulse

sequences. First, Eq. (8) reduces to

$$(-1)^{L} \sum_{j=0}^{L} A_{L}^{j}(x) \Phi_{L}^{j}(\vec{\varphi}) = e^{ix\gamma} + O(x^{n+1})$$
 (9)

by summing its even and odd parts, justified by noting $A_L^j(\theta_0/2+x) \to (-1)^L A_L^j(x)$; hence x occurs only in even (odd) powers for j even (odd). By matching coefficients of powers of x, this represents 2(n+1) real constraints. Second, the x^0 terms in Eq. (9) match if and only if L is even. Assuming this, 2n constraints remain. Third, we arrive at our most important result by transforming Eq. (9) with the substitution $x \to i \tanh^{-1}(y)$. This eliminates trigonometric and exponential functions from Eq. (9), and (assuming L is even) leaves

$$[(1-y)(1+y)]^{-L/2} \sum_{j=1}^{L} y^j \Phi_L^j(\vec{\varphi}) = \left[\frac{1-y}{1+y} \right]^{\gamma/2}.$$
 (10)

Upon rearrangement, this is a generating function for equations that the phase sums $\Phi_L^j(\vec{\varphi})$ must satisfy. The functions $f_L^j(\gamma)$ generated by $\sum_{j=0}^{\infty} f_L^j(\gamma) y^j = (1+y)^{(L-\gamma)/2} (1-y)^{(L+\gamma)/2}$ are, in fact, real polynomials in γ of degree j which generalize those of Mittag-Leffler [17]. We can now write

$$\Phi_I^j(\vec{\varphi}) = f_I^j(\gamma), \quad 0 < j \le n, \tag{11}$$

$$f_L^j(\gamma) = \sum_{k=0}^j (-1)^k {T \choose k} {L-T \choose j-k}, \quad T \equiv \frac{1}{2} (\gamma + L). \quad (12)$$

Equation (11) is, in our opinion, the simplest and most useful representation of the nonlinear (in $\vec{\varphi}$) constraints that form the basis for our solutions. We provide in Table II their explicit expansions for L=2,4,6 as examples.

TABLE II. Shown are explicit examples of the phase sums $\Phi_j^L(\vec{\varphi})$ defined in Eq. (7) and the polynomials $f_j^L(\gamma)$ defined in Eq. (12). From the definitions, we see $\Phi_0^L(\vec{\varphi}) = f_0^L = 1$, not shown in the table. For L = 2,4, when the expressions are still relatively short, we show $\Phi_j^L(\vec{\varphi})$ and $f_j^L(\gamma)$ for all j for completeness, although only those with $j \leqslant L/2$ are used in solving for our pulse sequences. Note that $\Phi_j^L(\vec{\varphi})$ is a sum of $\binom{L}{i}$ complex unit vectors and $f_j^L(\gamma)$ is a j-degree polynomial with either even or odd symmetry.

L	j	$\Phi^L_j(ec{arphi})$	$f_j^L(\gamma)$
2	1	$e^{iarphi_1}+e^{iarphi_2}$	$-\gamma$
	2	$e^{i(arphi_1-arphi_2)}$	$\frac{1}{2}\gamma^2-1$
4	1	$e^{iarphi_1}+e^{iarphi_2}+e^{iarphi_3}+e^{iarphi_4}$	$-\gamma$
	2	$e^{i(\varphi_1 - \varphi_2)} + e^{i(\varphi_1 - \varphi_3)} + e^{i(\varphi_1 - \varphi_4)} + e^{i(\varphi_2 - \varphi_3)} + e^{i(\varphi_2 - \varphi_4)} + e^{i(\varphi_3 - \varphi_4)}$	$\frac{1}{2}\gamma^{2} - 2$
	3	$e^{i(\varphi_1-\varphi_2+\varphi_3)} + e^{i(\varphi_1-\varphi_2+\varphi_4)} + e^{i(\varphi_1-\varphi_3+\varphi_4)} + e^{i(\varphi_2-\varphi_3+\varphi_4)}$	$-\frac{1}{6}\gamma^{3} + \frac{5}{3}\gamma$
	4	$e^{i(arphi_1-arphi_2+arphi_3-arphi_4)}$	$\frac{1}{24}\gamma^4 - \frac{2}{3}\gamma^2 +$
6	1	$e^{iarphi_1} + e^{iarphi_2} + e^{iarphi_3} + e^{iarphi_4} + e^{iarphi_5} + e^{iarphi_6}$	$-\gamma$
	2	$e^{i(\varphi_1-\varphi_2)} + e^{i(\varphi_1-\varphi_3)} + e^{i(\varphi_1-\varphi_4)} + e^{i(\varphi_1-\varphi_5)} + e^{i(\varphi_1-\varphi_6)} + e^{i(\varphi_2-\varphi_3)} + e^{i(\varphi_2-\varphi_4)} + e^{i(\varphi_2-\varphi_5)} + e^{i(\varphi_2-\varphi_6)} + e^{i(\varphi_3-\varphi_4)} + e^{i(\varphi_3-\varphi_5)} + e^{i(\varphi_3-\varphi_6)} + e^{i(\varphi_4-\varphi_5)} + e^{i(\varphi_4-\varphi_6)} + e^{i(\varphi_5-\varphi_6)}$	$\frac{1}{2}\gamma^2 - 3$
	3	$e^{i(\varphi_1-\varphi_2+\varphi_3)} + e^{i(\varphi_1-\varphi_2+\varphi_4)} + e^{i(\varphi_1-\varphi_2+\varphi_5)} + e^{i(\varphi_1-\varphi_2+\varphi_6)} + e^{i(\varphi_1-\varphi_3+\varphi_4)} + e^{i(\varphi_1-\varphi_3+\varphi_5)} + e^{i(\varphi_1-\varphi_3+\varphi_6)} + e^{i(\varphi_1-\varphi_4+\varphi_5)} + e^{i(\varphi_1-\varphi_4+\varphi_6)} + e^{i(\varphi_1-\varphi_5+\varphi_6)} + e^{i(\varphi_2-\varphi_3+\varphi_4)} + e^{i(\varphi_2-\varphi_3+\varphi_6)} + e^{i(\varphi_2-\varphi_3+\varphi_6)} + e^{i(\varphi_2-\varphi_4+\varphi_5)} + e^{i(\varphi_2-\varphi_4+\varphi_6)} + e^{i(\varphi_2-\varphi_4+\varphi_6)} + e^{i(\varphi_3-\varphi_4+\varphi_5)} + e^{i(\varphi_3-\varphi_4+\varphi_6)} + e^{i(\varphi_$	$-\tfrac{1}{6}\gamma^3 - \tfrac{8}{3}\gamma$

In our notation the leading error of an order n, even L, 2π -pulse sequence $S_{2\pi}$ has a simple form,

$$S_{2\pi} R_0[2x\gamma] = I - \left(f_L^{n+1}(\gamma) \hat{X}^{n+1} - \hat{\Phi}_L^{n+1}(\vec{\varphi}) \right) (-ix)^{n+1}.$$
(13)

Now, we recognize the operator on the right of Eq. (13) must be unitary. Thus, if a set $\vec{\varphi}$ satisfies Eq. (11) for 0 < j < k for any even integer k, Re[$\Phi_L^k(\vec{\varphi})$] = $f_L^k(\gamma)$ follows automatically. So we define $\mathcal{B}_{n,L}^{2\pi}$, the set of constraints resulting from applying $\theta_0 = 2\pi$ to $\mathcal{B}_{n,L}$, to consist of the n complex equations from Eq. (11), ignoring the real parts for even j.

$$\mathcal{B}_{n,L}^{2\pi} = \begin{cases} \operatorname{Re} \Phi_L^j(\vec{\varphi}) = f_L^j(\gamma), & j \text{ odd} \\ \operatorname{Im} \Phi_L^j(\vec{\varphi}) = 0, & \text{for all } j \end{cases}_{j=1,2,\dots,n}$$
 (14)

Thus, $|\mathcal{B}_{n,L}^{2\pi}| = \lceil 3n/2 \rceil$.

In fact, it is not difficult to place a lower bound on the pulse length L for a sequence correcting to order n using the framework we have so far. This is the first bound of its kind, and, given our solutions of the constraints to come in Sec. IV, it must be tight to a constant factor. Begin the argument by way of contradiction, letting n > L. In examining $\mathcal{B}_{n,L}^{2\pi}$, observe $\Phi_L^j(\vec{\varphi}) = 0$ for $L < j \leqslant n$, but $f_L^j(\gamma)$ is a real polynomial in γ of degree j. Hence $0 = \Phi_L^n(\vec{\varphi}) = f_L^n(\gamma)$ cannot be satisfied for arbitrary γ . Likewise, if L = n, then $1 = |\Phi_L^n(\vec{\varphi})| = |f_L^n(\gamma)|$ cannot be satisfied for arbitrary γ . Thus L > n is necessary.

C. Assuming phase angle symmetries

Some constraints in $\mathcal{B}_{n,L}^{2\pi}$ can be automatically satisfied if appropriate symmetries on the phase angle are imposed. A symmetry property of the phase sums, $\Phi_L^j(\vec{\varphi}) = [\Phi_L^j((-1)^j\vec{\varphi}_R)]^*$ with reversed phase angles $\vec{\varphi}_R = (\varphi_L, \varphi_{L-1}, \ldots, \varphi_1)$, motivates us to impose a palindromic (antipalindromic) symmetry on the phases, $\vec{\varphi} = +\vec{\varphi}_R$ ($\vec{\varphi} = -\vec{\varphi}_R$), so that $\mathrm{Im}[\Phi_L^j(\vec{\varphi})] = 0$ for even (odd) j. Removing these equations from $\mathcal{B}_{n,L}^{2\pi}$, we are left with the subset $\mathcal{B}_{n,L}^{\mathrm{PD}}$ ($\mathcal{B}_{n,L}^{\mathrm{AP}}$). By definition, $\varphi_k^{\mathrm{AP}} = -\varphi_{L-k+1}^{\mathrm{AP}}$ and $\varphi_k^{\mathrm{PD}} = \varphi_{L-k+1}^{\mathrm{PD}}$. In both cases, we have $|\mathcal{B}_{n,L}^{\mathrm{PD}}| = |\mathcal{B}_{n,L}^{\mathrm{AP}}| = n$ real constraints to be satisfied by $\lceil L/2 \rceil$ real variables $\vec{\varphi}^{\mathrm{AP}}$ or $\vec{\varphi}^{\mathrm{PD}}$. With what minimum L is this possible.

IV. SOLVING THE CONSTRAINTS

We now satisfy the constraints $\mathcal{B}_{n,L}^{\text{PD}}$ and $\mathcal{B}_{n,L}^{\text{AP}}$ with sequences of length exactly L=2n, using three different methods—analytical, perturbative, and numerical—and achieving the linear lower bound for 2π -pulse sequences. Our solutions are nonrecursive; a lower-order sequence never appears as part of an order n sequence. Table I summarizes our results labeled by PDn (APn) for the palindromes (antipalindromes), as well as "tower of power" (ToPn) sequences, a name inspired by their visual appearance in Fig. 1, which are special APn sequences essential to our perturbative proof that length-optimal arbitrary n corrections for nontrivial γ exist.

Sections IV A, IV B, and IV C detail, respectively, the analytical, perturbative, and numerical solution methods and the corresponding results.

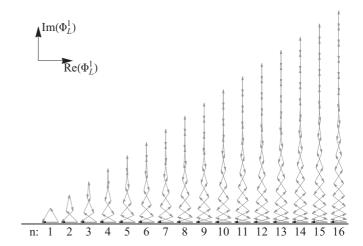


FIG. 1. Pulse sequences with $\theta_0=2\pi$ can be visualized with phasor diagrams as the first constraint equation; $\Phi_L^1(\vec{\varphi})=\sum_{k=1}^L e^{i\varphi_k}=f_L^1(\gamma)=-\gamma$ is a sum of phases. As examples, the phase angles $\vec{\varphi}_L|_{\gamma=1}=(\varphi_1,\ldots,\varphi_L)$ are plotted tip-to-tail to scale for ToPn sequences where L=2n and all arrows are of unit length. With $\vec{\varphi}$ chosen carefully, e.g., $\vec{\varphi}_2|_{\gamma=1}=(2\pi/3,-2\pi/3)$ or $\vec{\varphi}_4|_{\gamma=1}=(\cos^{-1}(\frac{\sqrt{10}-6}{4}),\cos^{-1}(1-\sqrt{\frac{5}{8}}),\ldots)$, higher-order constraints up to $\Phi_L^n(\vec{\varphi})=f_L^n(\gamma)$ are also satisfied, thus producing rotations correct to $O(\epsilon^n)$.

A. Closed-form solutions

We obtain closed-form solutions to $\mathcal{B}_{n,2n}^{AP}$ and $\mathcal{B}_{n,2n}^{PD}$ for $n \leq 3(4)$, presented in Table III, by the method of Gröbner bases [18], which we describe here. The sequences AP2, AP3, and PD4 are original, whereas AP1 and PD2 recover SK1 and PB₁ of Brown *et al.* [5] and Wimperis [10], respectively.

The key insight in solving the transcendental constraints $\mathcal{B}_{n,L}^{\text{AP}}$ and $\mathcal{B}_{n,L}^{\text{PD}}$ with γ as a free parameter is that any $\mathcal{B}_{n,L}$ is equivalent to systems of multivariate polynomial equations \mathcal{F} , for which powerful algorithmic methods of solution are known [19,20]. This equivalence can be seen by introducing the Weierstrass substitution $tan(\varphi_k/2) = t_k$. Any \mathcal{F} with variables $t_1, \ldots, t_n \in \mathbb{C}$ that has a finite number of zeros is zero dimensional and has solutions that can always be represented in the form of a regular chain [21], that is, a finite triangular system of polynomials $\{h_1(t_1), h_2(t_1, t_2), \dots, h_n(t_1, \dots, t_n)\}$ obtained by taking appropriate linear combinations of elements of \mathcal{F} . Regular chains are easy to solve, as the first equation h_1 is a univariate polynomial in t_1 whose zeros can then be substituted into h_2 , thus converting it into a univariate polynomial in t_2 . Through recursive substitution, all t_k can be obtained in a straightforward manner.

Divining these appropriate linear combinations appears to be a formidable task, but surprisingly, they can be deterministically computed by applying algorithms such as Buchberger's algorithm [22] for computing the Gröbner basis [18] \mathcal{G} of \mathcal{F} . The basis \mathcal{G} is another system of polynomial equations that shares the same zeros as \mathcal{F} , in addition to certain desirable algebraic properties. For example, \mathcal{G} can readily decide the existence, number of, and location of complex zeros [23], and by choosing a lexicographic *term order*, \mathcal{G} is itself a regular chain [18]. The algorithm generalizes

TABLE III. Closed-form solutions of phase angles $\varphi_k = 2 \tan^{-1}(t_k)$ for pulse sequences correcting to order $O(\epsilon^n)$ represented as regular chains, computed by the method of Gröbner bases [18]. Since $h(t_1) = 0$ is a univariate polynomial of degree ≤ 4 in t_1^2 , it can be solved in closed form. The other t_k are obtained directly by substitution. NS2 generalizes AP2 and PD2 with one free parameter t_4 , which should be fixed before solving.

Sequence	n	Phase angle solutions $\varphi_k = 2 \tan^{-1}(t_k)$					
AP1	1	$h(t_1) = (2 + \gamma) - (2 - \gamma)t_1^2.$					
AP2	2	$h(t_1) = (\gamma + 2)^2 (\gamma + 4) + 2\gamma (\gamma^2 + 4)t_1^2 + (\gamma - 4)(\gamma - 2)^2 t_1^4,$ $t_2 = t_1 \frac{2-\gamma}{2+\gamma}.$					
PD2	2	$h(t_1) = (4 + \gamma) - (4 - \gamma)t_1^2,$ $t_2 = -t_1.$					
NS2	2	$h(t_1, t_4) = (\gamma + 2)^2 (\gamma + 4) + \gamma (\gamma^2 - 4) (t_1^2 + t_4^2) - 16\gamma t_1 t_4 + (\gamma - 4)(\gamma - 2)^2 t_1^2 t_4^2,$ $t_{2(3)} = t_{1(4)} \frac{(\gamma - 4)(\gamma - 2) - 8(1 - \gamma)(1 + t_{4(1)}^2)^{-1} - 8(1 + t_{1(4)}^2)^{-1}}{(4 - \gamma)(\gamma + 2) - 8(\gamma + 1)(1 + t_{4(1)}^2)^{-1} - 8(1 + t_{1(4)}^2)^{-1}}.$					
AP3	3	$h(t_1) = a_1(\gamma) + 4a_2(\gamma)t_1^2 + 6a_3(\gamma)t_1^4 + 4a_2(-\gamma)t_1^6 + a_1(-\gamma)t_1^8, \text{ where } $ $a_1(\gamma) = (\gamma - 2)(\gamma + 2)^3(\gamma + 4)^2(\gamma + 6)^2, $ $a_2(\gamma) = (\gamma + 2)(\gamma + 4)(\gamma + 6)(\gamma^5 - 32\gamma^3 + 96\gamma^2 + 256\gamma + 192), $ $a_3(\gamma) = \gamma^8 - 60\gamma^6 + 816\gamma^4 + 9152\gamma^2 - 9216, $ $t_2^2 = -\frac{96\gamma + 16(\gamma^2 - 3\gamma + 2)(t_1^2 + 1) + (\gamma - 4)(\gamma - 2)(\gamma + 2)(t_1^2 + 1)^2}{96\gamma + 16(\gamma^2 - 9\gamma + 2)(t_1^2 + 1) + (\gamma - 6)(\gamma - 4)(\gamma - 2)(t_1^2 + 1)^2}, $ $t_3^2 = -\frac{\gamma + 2}{\gamma - 2} \frac{16(\gamma + 1) + (\gamma - 4)(\gamma - 2)(t_1^2 + 1)}{16(\gamma - 1) + (\gamma - 6)(\gamma - 4)(t_1^2 + 1)}, \text{ where signs of } t_{2,3} \text{ chosen to satisfy } $ $0 = \frac{t_1}{t_1} \left(\frac{2 - s_1}{t_1} + 2 \frac{2 - s_2}{t_2} + 2 \frac{2 - s_3}{t_2} \right) + \frac{t_3}{t_2} \left(\frac{2 - s_2}{t_2} + 2 \frac{2 - s_3}{t_2} \right) + \frac{t_3}{t_2} \frac{2 - s_3}{t_2}, s_n = 1 + t_n^2. $					
PD4	4	$h(t_1) = a_1(-\gamma) + 3b_1(\gamma)t_1^2 - 3b_1(-\gamma)t_1^4 - a_1(\gamma)t_1^6,$ $t_n = t_1(-1)^{1+\lceil n/2 \rceil} \frac{b_n(\gamma) + a_n(\gamma)t_1^2}{a_n(-\gamma) + b_n(-\gamma)t_1^2}, n = 2, 3, 4,$ $a_n(\gamma) = (\gamma - 8)(\gamma - 4)(\gamma + 4) \begin{cases} -(\gamma - 8)(\gamma - 4)^2\gamma, & n = 1, \\ -3\gamma(\gamma^2 - 2\gamma + 4), & n = 2, \\ \gamma(\gamma + 2)(\gamma^2 - 6\gamma - 4), & n = 3, \\ \gamma - 4, & n = 4 \end{cases}$					
		$b_n(\gamma) = (8+\gamma) \begin{cases} \gamma^6 - 64\gamma^4 + 128\gamma^3 + 1024\gamma^2 + 1024\gamma - 1024, & n = 1, \\ 5\gamma^5 - 10\gamma^4 - 76\gamma^3 - 64\gamma^2 - 64\gamma + 128, & n = 2, \\ (\gamma^2 - 6\gamma - 4)(\gamma^4 + 2\gamma^3 - 48\gamma^2 - 32\gamma - 64), & n = 3, \\ \gamma^3 + 4\gamma^2 - 64\gamma + 32, & n = 4 \end{cases}$					

Gaussian elimination for systems of linear equations and finding the greatest common divisor of univariate polynomial equations to systems multivariate polynomial equations; the reader is referred to excellent resources for more information [18–20]. In Appendix A, we also present a brief overview of Gröbner bases and Buchberger's algorithm for calculating them, including hand-worked examples for AP1 and PD2, the results of which are part of Table III.

The regular chains for the remaining sequences AP2, AP3, and PD4 solved in Table III can be computed by optimized variants of Buchberger's algorithm [18] in MATHEMATICA. In each case, closed form is achieved since $h(t_1)$ is a univariate polynomial of at most quartic degree in t_1^2 , and the remaining variables $t_{2,...,n}$ are then given as functions of only t_1 . Only the real solutions, which exist for $|\gamma| \le 2\lfloor n/2 \rfloor + 2$, are physically meaningful. The utility of Gröbner bases for short sequences is clear, as is it highly unlikely that these solutions could have been arrived at by hand.

As a curiosity, we also present in Table III a closed-form solution for $\mathcal{B}_{2,4}^{2\pi}$, where *no* symmetry has been applied to the four-pulse sequence, denoted NS2. NS2 has one free parameter in the phase angles, which we arbitrarily choose to be t_4 . By fixing t_4 and solving for the remaining phase angles, one finds

that NS2 continuously deforms between PD2 and AP2 and hence generalizes them.

Could one solve $\mathcal{B}_{n,2n}^{2\pi}$ for arbitrary n by this method? Any arbitrary system of multivariate polynomials is guaranteed to have a Gröbner basis that can always be computed in a finite number of steps by Buchberger's algorithm [22]. Thus complex solutions to zero-dimensional $\mathcal{B}_{n,2n}^{2\pi}$ with the same number of equations as free parameters φ_k can always be found by this method in principle. However, the worst-case time complexity of computing \mathcal{G} for a system of n variables and total degree d scales as $O(d^{2^n})$ [24] and rapidly becomes unfeasible. Of greater concern, there is no guarantee that such solutions are real, representing physical phases φ_k .

We now prove that there exists real solutions to $\mathcal{B}_{n,2n}^{2\pi}$ over a continuous range of γ for arbitrary n, and show how this leads to an efficient constructive procedure for computing arbitrary angle sequences.

B. Perturbative solutions

We may solve $\mathcal{B}_{n,2n}^{\mathrm{AP}}$ and $\mathcal{B}_{n,2n}^{\mathrm{PD}}$ perturbatively. A well-known theorem of square Jacobian matrices states that, given a set of equations $\mathcal{H}(\vec{\varphi}) = \{H_1(\vec{\varphi}) = 0, \dots, H_n(\vec{\varphi}) = 0\}$, then the set

 $\mathcal{H}(\vec{\varphi})$ is locally invertible, or analytical, in the neighborhood about some root $\vec{\varphi}_0$ if and only if the determinant $\det(J)$ of its Jacobian matrix $J_{jk} = \partial_{\varphi_k} H_j(\vec{\varphi})|_{\vec{\varphi} = \vec{\varphi}_0}$ is nonzero. The set of interest for us is $\mathcal{B}_{n,L}$, for which this theorem says that one may always construct a perturbative expansion for $\vec{\varphi}$ over a continuous range γ about γ_0 given a valid starting point $(\vec{\varphi}_0, \gamma_0)$ satisfying $\mathcal{B}_{n,L}$ if and only if $\det(J) \neq 0$. So long as the Jacobian remains nonzero, one may extend such a solution beyond its neighborhood by analytic continuation.

However, for arbitrary n, what are these valid initial points $(\vec{\varphi}_0, \gamma_0)$? As we do not *a priori* know of solutions to $\mathcal{B}_{n,L}$ for arbitrary γ , such points must be found at some γ where the problem simplifies. Even then, the problem is nontrivial: for example, imposing phase angle symmetries forces $\varphi_k = m\frac{\pi}{2}$, $m \in \mathbb{Z}$ at $\gamma = 0$, but one can readily verify that many such solutions of this form to Eq. (11) suffer from $\det(J) = 0$. Using the closed-form solutions in Table III, one finds at $\gamma = 0$ that while the Jacobian of the PD2,4 sequences is zero, the AP1,3 sequences each have a solution with a nonzero Jacobian wherein $\varphi_k = \pi/2$, for $k \le n$. We now prove that this generalizes to arbitrary n, resulting in the special class of ToPn antipalindrome sequences with initial values,

$$\left. \vec{\varphi}_n^{\text{ToP}} \right|_{\gamma = 2b} = \begin{cases} \pi, & 1 \leqslant k \leqslant b \\ \pi/2, & b < k \leqslant n, \end{cases} \tag{15}$$

for $b = 0, 1, \dots \lfloor n/2 \rfloor$. Hence nontrivial real solutions to $\mathcal{B}_{n,2n}^{2\pi}$ exist for arbitrary n.

1. ToPn is analytical at $\gamma = 0 \forall n$

We first transform the function mapping for ToPn: $\mathcal{H}(\vec{\varphi})_j = \{ ^{\operatorname{Re} \Phi^j_{2n}(\vec{\varphi}), \ j \text{ odd}}_{\operatorname{Im} \Phi^j_{2n}(\vec{\varphi}), \ j \text{ even}} \rightarrow \{ ^{\Phi^j_{2n}(\vec{\varphi}), \ j \text{ odd}}_{-i \Phi^j_{2n}(\vec{\varphi}), \ j \text{ even}} \text{. This does not affect the magnitude of its Jacobian } J \text{ as } \operatorname{Re} \Phi^j_L(\vec{\varphi}) = \Phi^j_L(\vec{\varphi}) \text{ for odd } j \text{ due to antipalindromic symmetry, while for even } j \text{ the real part of Eq. (11) is automatically satisfied due to unitarity [see Eq. (13)].}$

The $\gamma=0$ solution to ToPn has a simple form $\varphi_k^{\text{ToP}}=\pi/2$, for $k\leqslant n$. With this solution, a straightforward, if tedious, manipulation of the phase sums shows that elements of the $n\times n$ Jacobian matrix satisfy the recurrence $J_{j+1,k+1}=J_{j+1,k}+J_{j,k+1}+J_{jk}$. The solution to this recurrence is best seen from a combinatorial standpoint. Consider the related puzzle—you begin at the northwest corner (1,1) of an $h\times k$ checkerboard and would like to reach the position (h,k), the southeast corner. You may move only south, southeast, or east at any given time, enforcing the recursion. If, additionally, your first move cannot be south, how many paths exist that achieve your goal? The solution is

$$W_{hk} = \sum_{r=0}^{h-1} {k-1 \choose h-r-1} {k+r-2 \choose r},$$
 (16)

since you may take any number of southerly steps r. If your first move is not restricted, the number of paths is $D_{hk} = \sum_{p=1}^h W_{pk} = \sum_{r=0}^{h-1} \binom{k-1}{h-r-1} \binom{k+r-1}{r}$. For later use, define an $n \times n$ matrix D with the elements D_{jk} .

We will now express J_{jk} in terms of the leftmost column J_{i1} . This is an extension of the path counting problem, in which

we may begin our walk to (j,k) from any leftmost location. Therefore,

$$J_{jk} = \sum_{h=1}^{j} J_{j-h+1,1} W_{hk}.$$
 (17)

Now notice that the determinant of J does not depend upon the leftmost column. Since $J_{jk} = J_{j1}W_{0k} + \cdots + J_{11}W_{jk}$ and $J_{11} = -2 \neq 0$, we can always subtract multiples of rows of J to obtain -2D. Thus, $\det(J) = (-2)^n \det(D)$.

We have reduced the problem to finding the determinant of D. We claim that D has LU decomposition,

$$D_{jk} = \sum_{h=1}^{n} {j-1 \choose h-1} U_{hk}, \quad U_{hk} \equiv 2^{h-1} {k-1 \choose h-1}. \quad (18)$$

This means that D_{jk} is the binomial transform of the Chebyshev triangle U_{hk} . This is proved by looking at the generating functions of D and U, namely,

$$\mathcal{D}(y,z) = \sum_{j,k=1}^{\infty} D_{jk} y^{j-1} z^{k-1} = \frac{1}{1 - (y + yz + z)}, \quad (19)$$

$$\mathcal{U}(y,z) = \sum_{j,k=1}^{\infty} U_{jk} y^{j-1} z^{k-1} = \frac{1}{1 - (1+2y)z}.$$
 (20)

These are related by $\mathcal{D}(y,z) = \frac{1}{1-y}\mathcal{U}(\frac{y}{1-y},z)$, which implies the binomial transform in Eq. (18). With the LU decomposition, one can immediately see that $\det(D) = \det(U) = 2^{n(n-1)/2}$, and we have $\det(J) = (-1)^n 2^{n(n+1)/2} \neq 0$. This concludes the proof that ToP*n* sequences exist for a continuous range of small target angles γ within the neighborhood of γ_0 all n.

C. Numerical solutions

Our demonstrations of real, arbitrary angle (γ) solutions for small order (n) and real, arbitrary order solutions for a continuous range of small angles inspires confidence that real solutions for larger γ at arbitrary n can always be found. Although proving this notion is difficult, the zeroth-order analytically continuable solutions provide, in principle, a means of obtaining arbitrary γ , arbitrary n sequences that are exponentially more efficient than a brute force search for solutions to Eq. (11). As long as a sequence for some γ has a nonzero Jacobian, its phase angles may be continuously deformed into another solution to Eq. (11) in the neighborhood of γ .

We detail this procedure for ToPn and PDn to obtain real optimal-length solutions over $\gamma \in [0,2]$ in Secs. IV C 1 and IV C 2. In Sec. IV C 3 and Appendix B, we provide for the convenience of the reader the solutions derived in this manner at common values of $\gamma = \{1, \frac{1}{2}, \frac{1}{4}\}$ up to n = 12.

1. Analytic continuation of ToPn

We plot the Jacobian of ToPn solutions obtained by analytic continuation as a function of target angle γ in Fig. 2. The zeroth-order ToPn solutions in Eq. (15) can be continued from $\gamma = 0$ to arbitrary γ , up to n = 7, as J is nonzero over the range $\gamma \in [0,2]$. For $8 \le n \le 10$, analytic continuation from $\gamma = 2$

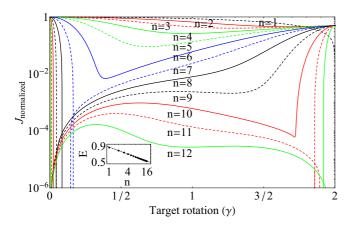


FIG. 2. (Color online) ToPn Jacobian $J_{\text{normalized}}$ normalized to 1 at $\gamma=0$ obtained by analytic continuation from zero-order solutions at $\gamma=0,1,2$ for $n\leqslant 12$. Analytic continuation to arbitrary angles from $\gamma=0$ is possible for n<7. From $7\leqslant n<11$, covering all γ requires continuation from $\gamma=1$ as well. For $11\leqslant n$, continuation from $\gamma=1$ is also necessary. Note that the n=10 curve appears have a discontinuous derivative near $\gamma=7/4$, but is, upon closer inspection, smoothly varying there. The inset plots the leading error $|f_{2n}^{n+1}-\Phi_{2n}^{n+1}|$ of ToPn at $\gamma=1$ up to n=16 together with the best-fit (dotted) $E=-O(\ln(n))$.

is required as well to cover the full range of γ , as J=0 at small γ , as seen in Fig. 2 (inset). For $11 \leqslant n$, we appear to encounter some difficulty as J=0 near $\gamma=0,2$. However, inspecting Fig. 1 suggests that an order n+1 sequence at $\gamma=1$ is approximated by appending to the order n sequence the phase angle $\varphi_{n+1}=\pi/2$. That is, with \circ representing the appending operation, $\vec{\varphi}_{n+1}^{\text{ToP}}|_{\gamma=1} \approx \vec{\varphi}_n^{\text{ToP}}|_{\gamma=1} \circ (\pi/2)$. This approximation is qualified by observing the monotonic decrease of the leading error $|f_{2n}^{n+1}-\Phi_{2n}^{n+1}|$ from Eq. (13) for ToPn sequences at $\gamma=1$. Thus, in the limit where $n\to\infty$, φ_n^{ToP} is a good initial guess for numerically finding the $\gamma=1$ ToP(n+1) root to $\mathcal{B}_{n,2n}^{\text{AP}}$. In this manner, we obtain the necessary zero-order solutions for continuation over all γ .

2. Analytic continuation of PDn

We would like to find for PDn a set of initial points $(\vec{\varphi}_0, \gamma_0)$ that are suitable for analytic continuation. These are provided by inspecting the PD2,4 closed-form solutions at $\gamma=2$ which are of the form $\vec{\varphi}_n^{\rm PD}=\frac{2\pi}{n+1}(1,-1,2,-2,\ldots,n/2,-n/2)$. We prove that these are arbitrary n solutions by noting that the sequence

$$V_n = \frac{2\pi}{n+1}(0,1,-1,2,-2,\ldots,n/2,-n/2), \qquad (21)$$

without palindrome symmetry applied, is in fact the length L=n+1 class B [4] sequence reported by Vitanov [15] that is optimal in the sense that for all odd j < L, $\Phi_L^j(\vec{\varphi}) = 0$ is satisfied. Thus, by applying palindrome symmetry and dropping the zeroth pulses, a length L=2n, $\gamma=2$ class A sequence is obtained. We conjecture that $\vec{\varphi}_n^{\text{PD}}$ is contained in

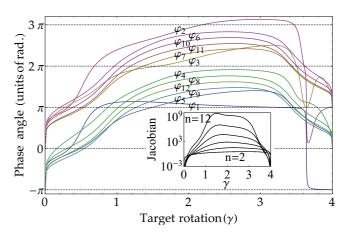


FIG. 3. (Color online) Phase angles φ_k as a function of target rotation $\gamma \in (0,4)$ for a length 24, order 12 palindromic PD12 sequence. This example demonstrates the method of analytic continuation from an exact solution $\vec{\varphi}_{12,6}^{PD}|_{\gamma=2} = \frac{12\pi}{13}(1,-1,2,-2,3,-3,4,-4,5,-5,6,-6)$, a procedure valid wherever the Jacobian of a sequence is nonzero, also plotted for $\vec{\varphi}_{n,n/2}^{PD}$ (inset). The dotted lines represent the expected continuation to the nonanalytic points $\gamma=0,4$.

the more general class of $\gamma = 2$ solutions:

$$\vec{\varphi}_{n,m}^{\text{PD}}\big|_{\gamma=2} = \frac{2m\pi}{n+1}(1,-1,2,-2,\dots,n/2,-n/2),$$
 (22)

where (n + 1, m) are coprime.

Arbitrary angle solutions to PDn, such as in Fig. 3, may be obtained by continuation from $\vec{\varphi}_{n,m}^{\text{PD}}|_{\gamma=2}$. Unlike the ToPn sequences, the Jacobian plotted in Fig. 3 (inset) for instances of $\vec{\varphi}_{n,n/2}^{\text{PD}}$ is nonzero over the entire range of γ up to n=12.

3. Brute-force numerical search

Another solution method is a direct numerical search for all possible real solutions to Eq. (11) for a given γ . This is by far the least efficient approach but allows one to obtain other classes of sequences that are not contained in $\vec{\varphi}_n^{\text{ToP}}$ and $\vec{\varphi}_{n,m}^{\text{PD}}$. Searching up to n=8 suggests that, after palindrome or antipalindrome symmetry has been applied, for every value of γ , $2^{\lceil n/2 \rceil}$ distinct real sequences exist. These sequences may also be analytically continued, and we provide in Table IV in Appendix B their phase angles at $\gamma=\{1,\frac{1}{2},\frac{1}{4}\}$ with which the interested reader may use to do so. In the table, we also provide a measure of the leading order error of each sequence κ , where κ is defined such that if the leading error from Eq. (13) is considered a small rotation, its amplitude is $(\kappa \epsilon \theta_0/2)^{n+1}$.

V. FURTHER EXTENSIONS

Some generalizations of the above results are possible. First, the sequence duration $L\theta_0/2\pi$ may be halved by considering $\theta_0=\pi$ sequences, with a significant decrease in the trace distance $\mathcal E$ from an ideal rotation as demonstrated in Fig. 4. This leads to similar Eqs. (9)–(13), but with replacements $\Phi_L^j(\vec\varphi) \to \Phi_L^{L-j}(\vec\psi)$ and $e^{ix\gamma} \to (-i)^L e^{ix\gamma}$. The x^0 term of the counterpart to Eq. (9) implies that $L \in 4\mathbf{Z}$.

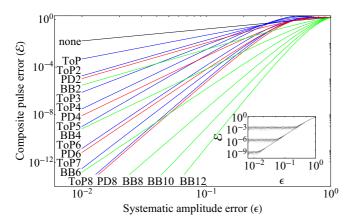


FIG. 4. (Color online) The trace distance of order n sequences U_T from an ideal π rotation, $\mathcal{E} = \|U_T - R_0[\pi]\|$, as a function of error ϵ . Included are $\theta_0 = 2\pi$ passband sequences ToPn (blue) and PDn (red), as well as the broadband $\theta_0 = \pi$ BBn (green) obtained by "toggling" PDn phase angles. Observe that the asymptotic gradient is $O(\epsilon^n)$, and that the number of 2π imperfect rotations is $n\theta_0/\pi$, given n is dramatically shorter than prior work by Brown et altharpoonup 1. [5]. The inset demonstrates the effects of experimental imprecisions by plotting \mathcal{E} for a population of BB4 sequences subject to a combined Gaussian distributed amplitude and phase errors with zero mean and standard deviation $\sigma = 10^{-3,-6,-9}$ from the top.

Second, while all 2π -pulse sequences are passband [10], broadband π -pulse sequences BBn with phase angles $\vec{\psi}$ can be obtained from palindrome sequences PDn with phase angles $\vec{\phi}$ through the "toggling" transformation [4,7], $\psi_k = -\sum_{h=1}^{k-1} (-1)^h \varphi_h + \sum_{h=k+1}^L (-1)^h \varphi_h$. Narrow-band π -pulse sequences are obtained simply by using the original phase angles $\vec{\phi}$.

Third, nonlinear systematic amplitude error suppression is also possible. Say the erroneous rotations making up sequence \mathcal{S} were instead $M_{\varphi}[\theta_0] = R_{\varphi}[\theta_0 + \epsilon E(\theta_0)]$ for any function E with $E(\theta_0) \neq 0$. Then the normalization $x \equiv \frac{1}{2}\epsilon E(\theta_0)$ and $\gamma \equiv E(\theta_T)/E(\theta_0)$ preserves Eqs. (8)–(13). Therefore, our pulse sequences correct nonlinear errors to $O(\epsilon^n)$, assuming the ratio $E(\theta_T)/E(\theta_0)$ is known.

Real applications involve additional systematic and random errors, assumed to be small, in both system and control. Examples of such include phase errors in φ as well as off-resonance errors [2]. These contribute to the primitive pulse $M_{\varphi}[\theta]$, a small effective rotation $R_{\vec{n}}[O(\delta)]$, $|\delta| \ll 1$, about some arbitrary axis \vec{n} . Note that interchanging the order $R_{\vec{n}}[O(\delta)] M_{\omega}[\theta] = M_{\omega}[\theta] R_{\vec{m}}[O(\delta)]$ only changes the rotation axis. Including uncorrected errors, our sequences become $S' = \prod_{i=1}^{L} R_{\vec{n}_i}[O(\delta_i)] M_{\varphi_i}[\theta_0] =$ $(\prod_{i=1}^{L} R_{\vec{m}_j}[O(\delta_j)])S$. Hence, the amplitude error suppression property of a pulse sequence is preserved, while amplifying other sources of error to $LO(\delta)$. If δ is purely random with variance σ^2 , this imposes a practical limit on the order n of error suppression $\sqrt{L}O(\sigma) \sim O(\epsilon^n)$, beyond which no decrease in pulse error is obtainable. This further implies that our sequences can be implemented even with significant experimental imprecision, which corresponds to at most linearly accumulating errors that simply level off at small ϵ , as illustrated in Fig. 4 (inset). Thus, in general, efficient sequences amplify uncorrected sources of error by at most L = poly(n).

Since amplitude error suppression to order n is only effective when other errors are small $|\delta| \sim |\epsilon|^n$, the simultaneous suppression of multiple sources of systematic errors is highly desirable. This can often be achieved by sequence concatenation [2,25]. For example, let $AMn = \prod_{k=1}^{L} M_{\varphi_k}[\theta_k]$ denote any sequence that suppresses systematic amplitude errors to order n. Noting that a $\theta = 2\pi$ rotation is already robust to off-resonance errors [25], replacing each $\theta \neq 2\pi$ pulse in AMn by a CORPSE sequence [9] results in a concatenated sequence suppressing off-resonance errors to first order and amplitude errors to order n. This follows from the observation that the CORPSE sequence approximating $R_{\varphi_T}[\theta_T]$ yields the rotation $M_{\varphi_T}[\theta_T] \cdot R_{\vec{n}}[O(\delta^2 + \delta \epsilon)]$ for some axis \vec{n} , where δ is the order of the off-resonance error [25]. This property is special to CORPSE, because the decomposition into $M_{\varphi_T}[\theta_T]$ times a small error will not generally occur.

VI. COMPARISON WITH PRIOR ART

There are very few sequences in the literature that allow for corrected arbitrary angle rotations [criterion (4) from our Introduction]. The classic examples SCROFULOUS [9] and the PB sequences [10] are only correct to n = 1,2, respectively. The work of Brown *et al.* [5] reports sequences SKn for arbitrary n, but with a length scaling of $L = O(n^3)$. Thus, for any given n, the class of optimal L = 2n sequences PDn, APn correct to same order as SKn, but with significantly fewer pulses, as seen by comparing their trace errors in Fig. 4 with those of Brown [5].

Therefore, for a more comprehensive comparison with prior art, we limit ourselves to inverting sequences so that $\theta_0 = \theta_T = \pi$ and the initial state is always $|0\rangle$. This corresponds to toggling the $\gamma = 1$ sequences from Appendix B and also to relaxing criteria (3) and (4) from our Introduction. For consistency of notion, the pulse length L will now include the zeroth pulse.

First, we note that the leading arbitrary accuracy $\theta_T = \pi$ class A sequences in the literature are derived by recursive nesting [8]. Given some order n_b base sequence comprised of L_b π pulses, one performs a nesting procedure [7] k times to obtain an order $n = (n_b + 1)^k - 1$ sequence with length $L = L_b^k$, corresponding to a length scaling of $L = (n+1)^{\log L_b/\log(n_b+1)}$. Wimperis' BB₁ sequence [10] with $L_b = 5$, $n_b = 2$ is often used as the base sequence, resulting in the F_r class of length $L = 5^r$, order $n = 3^r - 1$ sequences [8], with a asymptotic scaling of $L = O(n^{1.47})$, slightly worse than that of the L = 2n + 1 of BBn sequences reported here. Furthermore, our BBn sequences at $\gamma = 1$ can themselves be nested after rearranging in the toggled $\theta_0 = 2\pi$ frame such that the zeroth pulse is in the middle:

$$S_{2\pi} M_0 = M_{\varphi_1} M_{\varphi_2} \dots M_{\varphi_2} M_{\varphi_1} M_0 = R_0[2\pi] + O(\epsilon^{n+1})$$

= $M_{\varphi_n} \dots M_{\varphi_1} M_0 M_{\varphi_1} \dots M_{\varphi_n} + O(\epsilon^{n+1}).$ (23)

For example, if BB12 is used as the base sequence, $L = O(n^{1.25})$ is achieved. In the limit of large $L_b = 2n_b + 1$, L asymptotically approaches O(n).

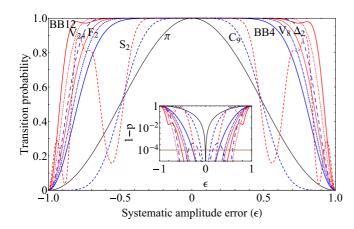


FIG. 5. (Color online) Comparison of the transition probability $p=|\langle 1|U_T|0\rangle|^2$ with amplitude error ϵ between a variety of L=9 (blue) and L=25 (red) pulse sequences implementing π rotations. Solid (dashed) lines represent class A(B) sequences. Included are the BB4,12 of this work, $V_{8,24}$ [Eq. (21)], Δ_2 [26], C_9 [27], F_2 [8], and S_2 [3]. In the inset, 1-p is plotted on a logarithmic scale.

Unlike class A sequences, class B sequences are less interesting from the context of quantum computing as they require specific initial states. The trace error metric is inapplicable, but one can nevertheless plot the transition probability $|\langle 1|U_T|0\rangle|^2$ as a function of ϵ for $\theta_T = \pi$ rotations acting on the ground state as in Fig. 5. As our highest-order closed-form sequence BB4 specialized to $\theta_T = \pi$ has L = 9, we perform a comparison with L = 9 inverting sequences in the literature: V_8 by Vitanov [15], $\Delta_2 = \frac{\pi}{6}(0,3,0,4,7,4,0,3,0)$ by Shaka and Freeman [26], and $C_9 = \frac{\pi}{12}(0,1,12,11,18,11,12,1,0)$ by Cho et al. [27]. Other than ours, there do not appear to be any L = 9 class A sequences. We also include a few L=25 sequences: the class A BB12 and F_2 [8,10], and the inverting sequences V_{24} and S_2 = $\frac{\pi}{3}(0,0,2,1,2,0,0,2,1,2,2,2,4,3,4,1,1,3,2,3,2,2,4,3,4)$ [3]. In this notation, $\theta_0 = \pi$ and, for example, C_9 is implemented by $M_0 M_{\frac{\pi}{12}} M_{\pi}$ We see that even against the specialized inverting sequences of similar length, BBn compares favorably. At the 10^{-4} quantum error threshold, the width of BBn of about $\epsilon = 0$ is only outperformed by the optimal inverting sequences V_n [15].

VII. CONCLUSION

The study of pulse sequences is a broad discipline, covering many different pulse shapes, error models, and computational methods, and we have dealt here with only a small portion of this breadth. However, our algebraic approach to amplitude errors provides a major characterization of this important and ubiquitous case. The constraints in Eq. (14) are both necessary and sufficient for any 2π -pulse sequence of length L correcting to order n. Using these constraints, we were able to find sequences in closed form beyond any order previously known analytically. From our 2π -pulse passband sequences, we also demonstrated that it is simple to obtain π -pulse broadband and π -pulse narrow-band sequences of the same optimal length.

In the Introduction, we proposed four criteria for a pulse sequence to be useful to quantum computation: (1) It exists at all orders of correction n. (2) It has efficient length L =

 $O(\operatorname{poly}(n))$. (3) It can operate as intended on any initial state. (4) It exists for all target rotations. We can evaluate our results with respect to these criteria. We proved that our $\operatorname{ToP} n$ and $\operatorname{PD} n$ sequences satisfy (1), (2), and (3) and provided evidence that (4) holds for them as well by studying the Jacobian of the phase sums and proving that $\operatorname{ToP} n$ solutions do indeed exist for a range of target rotations around the identity. At the same time, our numerical results indicate that conditions (2), (3), and (4) hold up to order n=12 with L=2n. This is a strong start to a complete proof that all four criteria are satisfiable for amplitude error correcting sequences. In fact, even a weaker proof that criteria (1), (2), and (3) hold at target angles π and $\pi/4$ would be valuable, as then pulse sequences for the Hadamard and T gate, together sufficient for universal single-qubit computation [1], could be implemented.

A very natural extension of our approach is to find constraints for sequences correcting other systematic control errors, for instance, arbitrary system drifts and off-resonance errors. As an example, Uhrig's [14] algebraic approach to dephasing errors might also be amenable to generalization. An anticipated problem in these other cases might be the nonpolynomial nature of such constraint equations, meaning more difficult analytic and numeric methods might be needed to find solutions.

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APPENDIX A: GRÖBNER BASES AND BUCHBERGER'S ALGORITHM

As the method of Gröbner bases is likely not well known to expected readers, we believe that a demonstration of deriving the bases for AP1 and PD2 would be instructive. Generalization to AP3, PD4 is straightforward but more computationally intensive. To begin, we use a definition of the Gröbner basis that facilitates its computation [22]:

Definition 1: Given a term order, a set G is a Gröbner basis if and only if for all $f,g \in G$ the S-polynomial SPOL(f,g) by repeated reduction with respect to G can be brought to zero.

```
Function: BUCHBERGER'S ALGORITHM (\mathcal{F}).
   \mathcal{G} \leftarrow \mathcal{F}
                                           ⊳System of multivariate polynomials
    C \leftarrow \mathcal{G} \times \mathcal{G}
    while C \neq \emptyset do
        p \leftarrow (a,b) \in C
                                                                        >Chosen arbitrarily
        C \leftarrow C \setminus \{p\}
        h = RED(SPOL(a,b), \mathcal{G})
                                                                             Defined below
       if h \neq 0 then
           C \leftarrow C \cup (\mathcal{G} \times \{h\})
            \mathcal{G} \leftarrow \mathcal{G} \cup \{h\}
       end if
    end while
                                                                    \trianglerightGröbner basis of \mathcal{F}
    return \mathcal{G}
end Function
```

This definition follows from Buchberger's theorem and leads to his famous algorithm for computing a Gröbner basis of \mathcal{F} [22]:

where

 $RED(a, \mathcal{G}) = Remainder of a upon division by \mathcal{G}$ (reduction),

$$SPOL(a,b) = lcm(LPP(a), LPP(b)) \left(\frac{a}{LM(a)} - \frac{b}{LM(b)} \right),$$

lcm(a,b) = least common multiple of a,b,

LM(a) = leading monomial of a

with respect to some term order,

LPP(a) = LM(a) with coefficients dropped.

In what follows, we apply the Weierstrass substitution $\tan(\varphi_k/2) = t_k$ to $\mathcal{B}_{n,L}$, rearrange to obtain a polynomial system $\mathcal{W}_{n,L}$, and use the lexicographic monomial ordering $t_1 \prec_{lex} t_2 \prec_{lex} \cdots \prec_{lex} t_n$ [18] in computing a Gröbner basis.

1. Example: AP1 from $\mathcal{B}_{1,2}^{AP}$

 $\mathcal{B}_{1,2}^{\mathrm{AP}} \Rightarrow \mathcal{W}_{1,2}^{\mathrm{AP}} = \{t_1^2(1+\frac{\gamma}{2}) - (1-\frac{\gamma}{2})\}$. This example is trivial, as $\mathcal{W}_{1,2}^{\mathrm{AP}}$ is automatically a Gröbner basis \mathcal{G} following from Definition 1 as the S polynomial of an arbitrary polynomial with itself is 0. As \mathcal{G} generates $\mathcal{W}_{1,2}^{\mathrm{AP}}$, they share the same simultaneous roots. Solving \mathcal{G} , we obtain $t_1 = \pm \sqrt{\frac{2-\gamma}{2+\gamma}}$. Solving for $\varphi_1 = \cos^{-1}(\frac{\gamma}{2})$, we see that AP1 is the sequence SK1 [5].

2. Example: PD2 from $\mathcal{B}_{2,4}^{PD}$

 $\mathcal{B}_{2,4}^{\text{PD}}\Rightarrow\mathcal{W}_{2,4}^{\text{PD}}=\{t_1^2t_2^2(\gamma-4)+t_1^2\gamma+t_2^2\gamma+(4+\gamma),t_1^2t_2+t_1t_2^2+t_1+t_2\}.$ In rearranging, we have introduced the complex roots $1+t_k^2=0$, which we shall have to remove later. One could solve $\mathcal{W}_{2,4}^{\text{PD}}$ by inspection, but we apply Buchberger's algorithm to demonstrate the algorithmic manner in which solutions may be derived. We perform the first iteration in detail and only state the computed basis element $h_{i,j}$ of succeeding iterations for brevity:

Input:
$$\mathcal{W}_{2,4}^{PD}$$
.

 $\mathcal{G} = \mathcal{W}_{2,4}^{PD}$, $C = \{(g_1, g_2)\}$,

Take the pair g_1, g_2 .

 $LM(g_1) = t_2^2 t_1^2 (\gamma - 4) = (\gamma - 4) LPP(g_1)$,

 $LM(g_2) = t_2^2 t_1 = LPP(g_2)$,

 $LCM(LPP(g_1), LPP(g_2)) = t_2^2 t_1^2$,

 $SPOL(g_1, g_2) = \frac{\gamma t_2^2}{\gamma - 4} - t_1^3 t_2 - t_1 t_2 + \frac{4t_1^2}{\gamma - 4} + \frac{\gamma + 4}{\gamma - 4}$,

 $h_{1,2} = RED(SPOL(g_1, g_2), \mathcal{G}) = SPOL(g_1, g_2)$,

 $\mathcal{G} = \mathcal{G} \cup \{h_{1,2}\}, C = \{(g_1, g_3), (g_2, g_3)\}$,

...

Output: $\mathcal{G} = \mathcal{W}_{2,1}^{PD} \cup \{h_{1,2}, h_{2,2}, h_{1,4}, h_{4,5}, h_{1,6}\}$

Output:
$$\mathcal{G} = \mathcal{W}_{2,4}^{PD} \cup \{h_{1,2}, h_{2,3}, h_{1,4}, h_{4,5}, h_{1,6}\},\$$

$$h_{2,3} = \frac{(\gamma - 4)t_2t_1^4 + 2(\gamma - 2)t_2t_1^2 + \gamma t_2 - 4t_1^3 - 4t_1}{\gamma},$$

$$\begin{split} h_{1,4} &= \frac{4t_1^3t_2 + 4t_1t_2 + \gamma t_1^4 + 2(\gamma + 2)t_1^2 + \gamma + 4}{\gamma - 4}, \\ h_{4,5} &= \frac{4\gamma t_2t_1^2 + 4\gamma t_2 - (\gamma - 4)\gamma t_1^5 - 2(\gamma - 2)\gamma t_1^3 - \gamma^2 t_1}{4(4 - \gamma)}, \\ h_{1,6} &= \frac{(\gamma - 4)t_1^6 + (3\gamma - 4)t_1^4 + (3\gamma + 4)t_1^2 + \gamma + 4}{4}, \\ \text{all other } h_{i,j} &= 0. \end{split}$$

Note that the last term $h_{1,6}$ is univariate in t_1 , as expected from a regular chain. We have chosen pairs from C so as to minimize the output, but any arbitrary choice will eventually terminate. However, \mathcal{G} from Eq. (A1) still contains more elements than is necessary to generate $\mathcal{W}_{2,4}^{PD}$. We can deterministically compute from \mathcal{G} a unique minimal, or *reduced*, Gröbner basis \mathcal{G}_R up to constant factors by repeating $\mathcal{G} \leftarrow (\mathcal{G} - \{g\}) \cup \{RED(g,\mathcal{G} - \{g\})\}\ \forall g \in \mathcal{G}$ until the process converges [18]:

$$G_R = \{ (t_1^2 + 1)^2 [4 + \gamma - (4 - \gamma)t_1^2],$$

$$(t_1^2 + 1) [(\gamma - 4)t_1^3 + \gamma t_1 - 4t_2],$$

$$4 + \gamma + (\gamma - 4)t_1^4 + 2(\gamma - 2)t_1^2 + 4t_2^2 \}.$$
 (A2)

The nonphysical zeros $\mathcal{M} = \{1 + t_1^2, 1 + t_2^2\}$ that were introduced earlier are now apparent and can be removed. We can deterministically compute from \mathcal{G}_R another Gröbner basis $\mathcal{G}_{\mathcal{Q}}$ with the same zeros sans \mathcal{M} by repeatedly computing the Gröbner basis of the ideal quotient $\mathcal{G}_{\mathcal{Q}} \to \langle \mathcal{G}_{\mathcal{Q}} \rangle : \langle \mathcal{M} \rangle$ until convergence, or *saturation* [18]. Finally, we obtain the simple triangular system

$$\mathcal{G}_O = \left\{ 4 + \gamma - (4 - \gamma)t_1^2, t_1 + t_2 \right\}. \tag{A3}$$

Solving for $\varphi_1 = \cos^{-1}\left(\frac{\gamma}{4}\right)$, $\varphi_2 = -\varphi_1$, we see that PD2 is the sequence PB₁ [10].

APPENDIX B: NUMERICAL SOLUTIONS

Here in Table IV, we provide phase angles for APn and PDn sequences at some common values of $\gamma \in \{1,\frac{1}{2},\frac{1}{4}\}$ up to n=12. Up to n=8, we provide the phase angles for all existing symmetric solutions at the selected values of γ . Sequences with the same subscript are related by analytic continuation and are sorted by their leading error κ at $\gamma=\frac{1}{2}$. Recall from Sec. IV C 3 that κ is defined so that $(\kappa\epsilon\theta_0/2)^{n+1}$ is the rotation angle of the leading error in Eq. (13) when it is considered as a small rotation. In the table, the bolded APn are ToPn sequences, and all PD n_1 sequences are obtained by analytic continuation from $\vec{\varphi}_{n,n/2}^{PD}|_{\gamma=2}$. Note that $\varphi_k^{AP}=-\varphi_{L-k+1}^{AP}$ and $\varphi_k^{PD}=\varphi_{L-k+1}^{PD}$. Also, for each sequence $\vec{\varphi}$ listed here, there is a sequence $-\vec{\varphi}$ with the same leading error.

TABLE IV. APn and PDn solutions at correction orders n = 1, 2, ..., 12 for target angles $\gamma = 1/4, 1/2, 1$.

γ	n = 1	$arphi_1$	К	n=2	φ_1	φ_2	!	К	n = 3	φ_1	φ_2	φ_3	К
1	AP1 ₁	2.0944	40 1.3160	07 PD2 ₁ AP2 ₁	1.82348 2.35949			1.16499 1.16499	AP3 ₁ AP3 ₂	0.74570 2.51806	-2.11099 1.15532	2.37504 1.66273	1.10279 1.10279
$\frac{1}{2}$	AP1 ₁	1.8234	48 0.9840	00 PD2 ₁ AP2 ₁	1.69612 1.95071			0.88856 1.05957	AP3 ₁ AP3 ₂	0.87848 2.03611	-1.93555 1.29441	2.13129 1.64504	0.93360 1.12057
$\frac{1}{4}$	AP1 ₁	1.6961	12 0.7043	PD2 ₁ AP2 ₁	1.63334 1.75891			0.69647 0.86557	AP3 ₁ AP3 ₂	0.98173 1.80090	-1.85668 1.42667	1.98076 1.61136	0.77869 0.98560
γ	n = 4	$arphi_1$	$arphi_2$	$arphi_3$	$arphi_4$	κ	n = 5	φ_1	$arphi_2$	$arphi_3$	$arphi_4$	$arphi_5$	κ
1	PD4 ₁ PD4 ₂ AP4 ₁ AP4₂	2.26950 1.38777 1.64767 2.62323	-1.76948 -0.59527 -1.97451 0.99049	-0.80579 2.74476 2.92461 1.79394	-2.19903	1.07009 1.07009 1.07009 1.07009	AP5 ₁ AP5 ₂ AP5 ₃ AP5 ₄	0.44569 2.19409	-2.57163 -1.40804 0.13026 0.86163	1.55593 5 –2.09113		2.50831 1.72591	1.05043 1.05043 1.05043 1.05043
$\frac{1}{2}$	PD4 ₁ PD4 ₂ AP4 ₁ AP4₂	2.30661 1.47070 1.05532 2.10426		-0.47998 -2.96678 3.06746 1.80109	0.26240	0.88941 0.89673 0.90343 1.15341	AP5 ₁ AP5 ₂ AP5 ₃ AP5 ₄	0.60281 1.78432	-2.24227 -1.44347 0.20550 0.89078	7 1.4503 07 -2.2189	1 –2.28880 7 1.86132	2.29567 1.69801	0.91458 0.93241 1.05179 1.13467
$\frac{1}{4}$	-	2.45079 1.51926 0.68460 1.83302		-0.28079 -2.38182 3.11442 1.70166		0.76705 0.77110 0.77643 1.07442	AP5 ₁ AP5 ₂ AP5 ₃ AP5 ₄	0.73268 1.62724	-2.19365 -1.46554 0.36022 1.22698	1.41033 2 –2.23779	3 –2.18996 9 1.88279	2.15661 1.64971	0.80141 0.82226 0.95228 1.13458
γ	n	= 6	$arphi_1$		$arphi_2$	φ_3		$arphi_4$		$arphi_5$	φ_6	i	κ
1	F F F F	PD6 ₁ PD6 ₂ PD6 ₃ PD6 ₄ AP6 ₁ AP6 ₂ AP6 ₃	2.48390 2.24222 1.17370 0.38266 2.54809 1.92279 0.74467 2.75387	-2. -0. -3. 2. -3. -2.	63561 44350 19700 00356 02296 02711 15643 76025	-0.230 1.642 2.333 -2.244 -0.586 -0.188 2.730 2.04	224 177 117 625 830 082	2.032 -1.082 -0.999 2.230 0.736 -1.614 2.723 1.353	266 025 067 067 027 142	2.74686 -1.76926 3.05926 -1.43621 2.99308 2.05848 -0.85131 1.63574	-0.71 0.81 -2.37 0.84 1.38 1.19 1.05	242 762 607 890 384 986	1.03757 1.03757 1.03757 1.03757 1.03757 1.03757 1.03757
$\frac{1}{2}$	F F <i>F</i> <i>A</i>	PD6 ₁ PD6 ₂ PD6 ₃ PD6 ₄ AP6 ₁ AP6 ₂ AP6 ₃ AP6₄	2.71338 2.40846 1.35661 0.34769 2.11291 1.56304 0.97792 2.26941	-1. 0. -2. 2. -1.	89153 52806 50760 51801 26524 92131 74876 51330	0.349 3.022 2.849 2.110 -0.553 -0.500 2.493 2.352	225 949 929 309 959 396	2.885 0.143 -0.349 -1.905 0.482 -1.717 2.839	373 944 548 262 787	-2.77240 -1.19151 -2.58938 1.90034 2.87662 1.85652 -0.72251 1.64389	0.12 1.48 -2.05 -0.66 1.43 1.29 1.19	740 672 420 607 826 155	0.89347 0.89824 0.90808 0.91063 0.95146 0.97397 0.98162 0.99856
$\frac{1}{4}$	F F <i>F</i> <i>A</i>	PD6 ₁ PD6 ₂ PD6 ₃ PD6 ₄ AP6 ₁ AP6 ₂ AP6 ₃	3.12733 2.66911 1.46125 0.61658 1.82740 1.43968 1.09750 1.88994	-0. 1. -2. 2. -1.	27154 96400 00480 23483 77886 68323 53786 07769	0.71° -2.590 2.91° 2.05° -0.118 -0.674 2.385 2.113	575 140 399 320 496 588	-2.883 0.769 -0.242 -1.709 0.335 -1.735 2.889 1.238	929 281 970 508 584 953	-2.34449 -0.86098 -2.10817 2.20472 2.95974 1.74153 -0.61467 1.68167	0.60 1.89 -1.82 -0.51 1.46 1.37 1.30	620 942 392 074 503 211	0.80563 0.80770 0.81322 0.81490 0.84058 0.87521 0.88747 1.14696

TABLE IV. (Continued.)

,	n = 7	$arphi_1$	$arphi_2$	φ_3	3	φ_4	$arphi_5$	φ_6	$arphi_7$	κ
	$AP7_1$	2.13314	1.81689			35267	2.63930	0.24057	2.09463	1.028
	$AP7_2$	0.46860	-1.07834			48204	1.99013	-2.42873	2.39994	0.936
	$AP7_3$	1.53566	-1.94880	-3.10	1.0873	06287	-1.02408	0.63518	-2.93056	1.028
	$AP7_4$	2.79719	0.67921	2.15	156 1.	25324	1.70211	1.53591	1.57511	1.028
	$AP7_5$	2.67233	1.22852	-0.41	410 2.	09290	3.01643	0.79095	1.64235	1.028
	$AP7_6$	2.38920	2.56740	-1.87	487 1.	51487	-0.58086	0.89490	1.71973	1.028
	$AP7_7$	1.90502	-0.25506	-0.74	525 2.	18555	-2.05722	2.23559	1.78586	1.028
	$AP7_8$	2.56602	0.68912	0.58	-2.	14580	2.36676	1.54736	1.60313	1.028
	$AP7_1$	1.71155	2.56416			38344	2.74047	0.11831	2.03574	0.922
	$AP7_2$	0.46860	-1.07834			48204	1.99013	-2.42873	2.39994	0.936
	$AP7_3$	0.62501	-2.74449				-0.93150	0.79294	-2.69744	0.943
	$AP7_4$	2.33062	0.29389			28505	1.66008	1.54990	1.57324	0.946
	$AP7_5$	2.20315	1.13325	-0.64	167 2.	00550	2.99463	0.94849	1.62714	1.010
	$AP7_6$	1.92827	2.68589	-1.88	707 1.	50932	-0.51984	1.04819	1.69068	1.019
	$AP7_7$	1.58492	-0.10094	-0.98	974 1.	95733	-2.24104	2.21759	1.75068	1.024
	$AP7_8$	2.05179	0.74564	0.44	549 –2.	29705	2.45454	1.51269	1.61609	1.093
	$AP7_1$	1.44528	3.12551	0.18	508 –1.	47200	2.68933	-0.05344	2.01638	0.833
	$AP7_2$	0.59981	-1.15892	1.31	704 -1.	52059	1.89255	-2.33236	2.27275	0.850
	$AP7_3$	0.10274	3.04219	-2.58	618 1.	23233	-0.92136	0.93214	-2.51134	0.859
	$AP7_4$	1.97582	0.44006	2.74	911 1.	20237	1.65775	1.55408	1.57254	0.884
	AP7 ₅	1.90819	1.35140				-3.07859	1.01561	1.61079	0.898
	$AP7_6$	1.70670	2.60476				-0.56759	1.19451	1.65081	0.93
	$AP7_7$	1.48964	0.08106				-2.32236	2.18736	1.69126	0.94
	$AP7_8$	1.78462	1.06489			37747	2.24540	1.54878	1.60439	1.080
	n = 8	$arphi_1$	$arphi_2$	$arphi_3$	$arphi_4$	$arphi_5$	$arphi_6$	$arphi_7$	$arphi_8$	κ
	PD8 ₁	2.61153	-1.49208	0.15281	2.20741	-2.95859	-0.58046	-0.00546	2.86733	1.022
	$PD8_2$	2.50835		-1.12170	-3.08779	0.28846		1.27599	-2.22667	1.022
	PD8 ₃	2.18294	-2.97594	1.35331	-2.84491	0.16565		-2.25027	0.16552	1.022
	PD8 ₄	1.82404		-2.98208	-2.13737	-1.50007		0.11098	1.35471	1.022
	PD8 ₅	1.75477	-2.32150	0.12777	-1.18259	2.40815		0.30847	2.60823	1.022
	$PD8_6$	1.03767	-0.07640	1.75380	-1.00830	2.33601		-3.09970	-2.48577	1.02
	$PD8_7$	1.01659	-2.52557	2.89648	0.33374	-1.50338		-2.29474	0.09358	1.02
	$PD8_8$	1.74334		-2.39342	-0.72709	2.05671		2.48855	-1.09561	1.02
	$AP8_1$	0.83305	-1.73107	1.95053	-0.72709 -2.32641	2.89702		0.38560	0.26672	1.02
		1.61633		-1.51594	1.42758	0.18349		2.76220	0.28672	1.02
	AP8 ₂			1.75303						
	$AP8_3$	2.27849	-2.00209		-0.08067	0.27972		-2.81981	0.76531	1.02
	AP8 ₄	2.83177	0.61330	2.24038	1.15822	1.77967	1.49245	1.58945	1.56873	1.02
	AP8 ₅	2.35733		-2.82410	-0.44884	-1.30394		1.60198	1.48100	1.02
	AP8 ₆	2.13392		-2.41837	1.70116	2.70715		1.07784	1.46294	1.02
	AP8 ₇	2.59142	1.60949	0.74701	-2.11111	-0.81043 2.13587		2.01457	1.51056	1.02
	AP8 ₈	2.74925	0.90733	0.84501	-2.90598			1.93876	1.54157	1.02
	$PD8_1$ $PD8_2$	3.09896 2.89767	-0.40393 -0.75419	1.07238 0.02350	-2.81247 -0.44104	-1.81742 2.77895		1.25751 2.36633	-2.09357 -1.07052	0.89 0.90
	$PD8_3$	2.44672	-1.76010	2.76238	-2.67594	0.42718		-1.48585	0.99119	0.90
	$PD8_4$	2.15221		-1.28384	-2.07394 -1.48037	-0.86524		1.44250	2.04352	0.90
	PD8 ₅	1.79499	-2.20241	0.33989	-0.13113	-3.09034		0.64086	3.07444	0.90
	PD8 ₆	1.79499	0.59213	2.03022	-0.13113 -0.76950	2.47232		-2.48749	-2.13266	0.91
	PD8 ₇	0.21511	3.01940	3.12108	0.36417	-1.25531		-2.60242	-0.03202	0.92
	$PD8_8$	0.97746		-2.19087 1.70622	-1.03016	1.93847		2.35917	-1.11989	0.92
	AP8 ₁	0.61590	-1.54138	1.79633	-2.64106 0.70542	3.03586		0.25918	0.20325	0.92
	AP8 ₂	1.07682		-2.26979	0.70542	0.23122		2.88367	0.36618	0.92
	$AP8_3$	1.69369	-1.97187	2.20200	-0.14932	0.17567		-2.78478	0.67753	0.92
	AP8 ₄	2.36536	0.18413	2.37615	1.28246	1.68628		1.58055	1.56971	0.95
	A DO	1.88735	1.26877	-2.67285	-0.31979	-1.39755	2.12492	1.60375	1.47723	1.05
	AP8 ₅									
	AP8 ₅ AP8 ₆ AP8 ₇	1.73150 2.09427		-2.55773 0.60461	1.79669 -2.17505	2.76172 -0.63648	-1.23074	1.08662 1.98533	1.45651 1.50975	1.059 1.060

TABLE IV. (Continued.)

γ	n = 8	φ_1	φ_2	φ_3	φ_4	φ_5	$arphi_6$	$arphi_7$	$arphi_8$	κ
$\frac{1}{4}$	PD8 ₁	2.36533	-0.59348	-1.59387	2.11324	1.26477	-1.39671	-1.84750	1.47308	0.83193
4	PD8 ₂	2.77667	-0.07740		-0.44784	2.63816		-2.91175		0.83255
	$PD8_3$	2.78074	-1.06236	-2.83926	-2.58960	0.52429		-1.06493	1.48454	0.83488
	$PD8_4$	2.42390	2.91231	-0.17275	-0.98092	-0.48142	-0.88408	2.29932	2.48783	0.83624
	PD8 ₅	1.90609	-1.94011	0.59178	0.66409	-2.37678	-1.59005	0.85358	-2.89166	0.83956
	$PD8_6$	1.42220	1.05360	1.96382	-0.77487	2.39390	-1.16804	-2.05033	-1.87167	0.84089
	$PD8_7$	0.15447	-2.65067	3.11263	-0.30804	1.12308	-2.27416	2.84794	0.16435	0.84309
	$PD8_8$	0.52355	-2.82738	-2.13271	-1.37376	1.67229	0.88411	2.21759	-1.19502	0.84360
	$AP8_1$	0.43759	-1.25361	1.97034	-2.83601	3.08638	-2.92195	0.17407	0.14459	0.84606
	$AP8_2$	0.74902	-2.91162	-2.58867	0.39895	0.21710	-2.70507	2.94458	0.29709	0.84732
	$AP8_3$	1.25748	-2.21044		-0.18148	0.13282		-2.79560	0.58885	0.85238
	$AP8_4$	1.99996	0.25518		1.27163	1.65234		1.57614		0.89121
	$AP8_5$	1.68405	1.39892	-2.66779	-0.14679	-1.40829		1.54453		1.00109
	$AP8_6$	1.58930	0.07907		1.94431	2.74375		1.13869		1.00433
	$AP8_7$	1.82114	1.97469		-2.07905	-0.55151		1.85296		0.99405
	$AP8_8$	1.93220	0.95664	1.13275	-2.40924	2.34418	0.46327	1.81690	1.55146	0.98427
γ	$arphi_1$	$arphi_2$	φ_3	φ_4 φ_5	$arphi_6$	$arphi_7$	φ_8 φ_9	$arphi_{10}$	φ_{11} φ_{12}	κ
1 A	AP9 2.8600	0.55880	2.31608	1.07164 1.86035	1.43542	1.61698 1	.56086 1.57179	9 –		1.01731
Pl	D10 ₁ 2.6967	0 -1.35876	0.44672	2.41011 -2.50642	-0.38506	0.55856 3	.03426 -2.7662	6 0.14621		1.01358
A	P10 2.8835	1 0.51305	2.38089	0.99425 1.93898	1.37051	1.65721 1	.54389 1.5760	5 1.57032		1.01358
				0.92559 2.01287			.51647 1.5863		1.57103 –	1.01065
Pl	D12 ₁ 2.7578	0 - 1.24040		2.61730 -2.13289					0.75161 -2.5895	6 1.00830
A	P12 2.9203	9 0.44071	2.48526	0.86478 2.08092	1.23560	1.76313 1	.47942 1.6044	7 1.56194	1.57226 1.5706	8 1.00830
$\frac{1}{2}$	AP9 2.3918:	5 0.10802	2.35997	1.26948 1.72330	1.49690	1.59735 1	.56486 1.5714	1 –		0.98012
	D10 ₁ 2.6366	3 -0.29344	-1.78826	2.13382 0.91161	-1.45848	-2.30168 1	.36696 0.93599	9 -2.02087		0.90538
A	P10 2.8835	0.51305	2.38089	0.99425 1.93898	1.37051	1.65721 1	.54389 1.5760	5 1.57032		1.00431
A	P11 2.9033	8 0.47416	2.43675	0.92559 2.01287	1.30273	1.70718 1	.51647 1.5863	1 1.56802	1.57103 –	1.02570
Pl	D12 ₁ 2.75780	0 -1.24040	0.68112	2.61730 -2.13289	-0.15967	1.03230 -3	.03873 -2.21969	9 0.33514	0.75161 -2.5895	6 0.91310
A	P12 2.9203	9 0.44071	2.48526	0.86478 2.08092	1.23560	1.76313 1	.47942 1.6044	7 1.56194	1.57226 1.5706	8 1.03977
$\frac{1}{4}$	AP9 2.0165	7 0.13293	2.66164	1.30204 1.66304	1.53099	1.58483 1	.56766 1.5711	2 –		0.92234
				1.28850 0.29208				4 –2.7124		0.85202
	P10 2.0305				1.50560		.56014 1.5730	0 1.57059		0.95542
	P11 2.0428								1.57094 –	0.98701
Pl	D12 ₁ 0.3791	8 -2.8705	-3.06176 -	-0.38647 0.58995		-2.34477 0	.14723 0.8115	0 –2.76097 –	2.41850 0.5735	3 0.86785
				1.31975 1.74694				4 1.56350	1.57206 1.57069	9 1.01568

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