

I. SYSTEM MODEL

Consider an IRS-aided multiuser SISO SWIPT system where the IRS not only assists the primal transmission but also retrieves Channel State Information (CSI) and harvests energy for its own operation. The single-antenna transmitter delivers information and power simultaneously, through the L -reflector IRS, to K single-antenna users over N orthogonal subbands. It is assumed that the IRS performs channel estimation in the first subframe and supports information and power transfer in the second subframe [1]. Due to the passive characteristics of IRS, we consider a Time-Division Duplexing (TDD) protocol where the CSI can be obtained by exploiting channel reciprocity. Perfect CSI is assumed at the AP and IRS to investigate the analytical upper-bound of the proposed system. The signals reflected by IRS for two and more times are assumed negligible and thus not considered. A quasi-static frequency-selective model is used for both the AP-user and AP-IRS-user links where the channels are assumed unchanged within each transmission frame. A superposition of multicarrier modulated and unmodulated waveforms, both transmitted on the same frequency bands, are designed adaptively to maximize the rate-energy tradeoff [2]. Note that although Frequency Selective Surface (FSS) has received much attention for wideband communications, active FSS requires RF-chains thus becomes prohibitive in IRS [3], [4]. Since passive FSS is not reconfigurable with fixed physical characteristics [5], we assume a frequency-flat IRS with same reflection coefficients for all subbands. Two practical receiver architectures proposed in [6], namely Time Switching (TS) and Power Splitting (PS), are investigated for the co-located information decoder and energy harvester. In the TS strategy, each transmission subframe is further divided into orthogonal data and energy slots with duration ratio $(1 - \alpha)$ and α respectively. Hence, the achievable rate-energy region can be obtained through a direct time sharing between wireless power transfer (WPT) with $\alpha = 1$ and wireless information transfer (WIT) with $\alpha = 0$. The adjustment of α has no impact on the transmit waveform and IRS elements design, since they are optimized individually in data and energy slots. In comparison, the PS scheme splits the received signal into data and energy streams with power ratio $(1 - \rho)$ and ρ . As ρ is coupled with waveform and IRS design, we investigate the rate-energy region by placing different rate constraints and optimizing waveform, PS ratio and reflectors accordingly. Perfect synchronization is assumed among the three parties in both scenario.

A. Transmitted Signal

1) *Modulated Information Waveform:* Assume the carrier are evenly spaced with equal bandwidth B_s such that the frequency of the n -th subband is $f_n = f_0 + (n - 1)\Delta f$ ($n = 1, \dots, N$). The information symbol $\tilde{x}_{I,k,n} \sim \mathcal{CN}(0, 1)$ of user k over subband n is a capacity-achieving i.i.d. Circular Symmetric Complex Gaussian (CSCG) variable with zero

mean and unit variance. Therefore, the baseband (BB) transmit information signal at subband n can be expressed as

$$x_{I,n}(t) = \sum_{k=1}^K w_{I,k,n} \tilde{x}_{I,k,n}(t) \quad (1)$$

where $w_{I,k,n}$ is the corresponding information weight. Let $\mathbf{W}_I = [\mathbf{w}_{I,1}, \dots, \mathbf{w}_{I,N}]^T \in \mathbb{C}^{N \times K}$ with $\mathbf{w}_{I,n} = [w_{I,1,n}, \dots, w_{I,K,n}]^T \in \mathbb{C}^{K \times 1}$. On top of this, the overall frequency weight is denoted as $w_{I,n} = \sum_k w_{I,k,n}$ and collected into $\mathbf{w}_I = [\mathbf{w}_{I,1}, \dots, \mathbf{w}_{I,N}]^T \in \mathbb{C}^{N \times 1}$. The RF transmit information waveform at time t writes as

$$\begin{aligned} x_I(t) &= \sum_{n=1}^N x_{I,n}(t) e^{j2\pi f_n t} \\ &= \sum_{k=1}^K \sum_{n=1}^N w_{I,k,n} \tilde{x}_{I,k,n}(t) e^{j2\pi f_n t} \end{aligned} \quad (2)$$

2) *Unmodulated Power Waveform:* As suggested in [2], [7], we use deterministic multisine waveform to boost the harvested energy. It has no randomness over time such that the baseband transmit power signal at subband n is

$$x_{P,n} = \sum_{k=1}^K w_{P,k,n} \quad (3)$$

where $w_{P,k,n}$ is the corresponding power weight. Let $\mathbf{W}_P = [\mathbf{w}_{P,1}, \dots, \mathbf{w}_{P,N}]^T \in \mathbb{C}^{N \times K}$ with $\mathbf{w}_{P,n} = [w_{P,1,n}, \dots, w_{P,K,n}]^T \in \mathbb{C}^{K \times 1}$. Similarly, the overall frequency weight is denoted as $w_{P,n} = \sum_k w_{P,k,n}$ and collected into $\mathbf{w}_P = [\mathbf{w}_{P,1}, \dots, \mathbf{w}_{P,N}]^T \in \mathbb{C}^{N \times 1}$. The transmit power signal at time t is given by

$$\begin{aligned} x_P(t) &= \sum_{n=1}^N x_{P,n} e^{j2\pi f_n t} \\ &= \sum_{k=1}^K \sum_{n=1}^N w_{P,k,n} e^{j2\pi f_n t} \end{aligned} \quad (4)$$

3) *Superposed Waveform:* At time t , a superposition of the information and power waveform at subband n writes as

$$x_n(t) = \sum_{k=1}^K (w_{I,k,n} \tilde{x}_{I,k,n}(t) + w_{P,k,n}) e^{j2\pi f_n t} \quad (5)$$

Hence, the transmitted RF signal is

$$x(t) = \sum_{k=1}^K \sum_{n=1}^N (w_{I,k,n} \tilde{x}_{I,k,n}(t) + w_{P,k,n}) e^{j2\pi f_n t} \quad (6)$$

B. Composite Channel Model

Denote the baseband equivalent channels from the AP to users, from the AP to the IRS, and from the IRS to users as $\mathbf{H}_D \in \mathbb{C}^{K \times N}$, $\mathbf{H}_I^T \in \mathbb{C}^{L \times N}$, and $\mathbf{H}_R \in \mathbb{C}^{K \times NL}$ respectively. At subband n , the frequency response of direct and reflective links write as $\mathbf{h}_{D,n} = [h_{D,1,n}, \dots, h_{D,K,n}]^T \in \mathbb{C}^{K \times 1}$ and $\mathbf{H}_{R,n} = [\mathbf{h}_{R,1,n}^T, \dots, \mathbf{h}_{R,K,n}^T]^T \in \mathbb{C}^{K \times L}$ with $\mathbf{h}_{R,k,n} = [h_{R,k,n,1}, \dots, h_{R,k,n,L}] \in \mathbb{C}^{1 \times L}$. Similarly, the n -th

incident channel is given by $\mathbf{h}_{I,n} = [h_{I,n,1}, \dots, h_{I,n,L}]^T \in \mathbb{C}^{L \times 1}$. To decouple the impact of IRS operation on the incident link, we construct a block-diagonal matrix $\mathbf{H}_I \triangleq \text{diag}\{\mathbf{h}_{I,1}, \dots, \mathbf{h}_{I,N}\} \in \mathbb{C}^{NL \times N}$. Element l of the IRS receives a superposed waveform through the multipath channel, then redistributes it by adjusting the amplitude reflection coefficient $\beta_l \in [0, 1]$ and phase shift $\theta_l \in [0, 2\pi)$. Each passive reflector absorbs a small portion $(1 - \beta_l)$ of the signal to support CSI decoding and impedance matching. On top of this, the IRS matrix per subband is constructed by collecting the reflection coefficients onto its diagonal entries as $\mathbf{\Theta}_0 = \text{diag}\{\beta_1 e^{j\theta_1}, \dots, \beta_L e^{j\theta_L}\} \in \mathbb{C}^{L \times L}$. Finally, the IRS matrix is formed by $\mathbf{\Theta} = \text{diag}\{\underbrace{\mathbf{\Theta}_0, \dots, \mathbf{\Theta}_0}_{N \text{ times}}\} \in \mathbb{C}^{NL \times NL}$.

The IRS-aided extra link can be modeled as a concatenation of the AP-IRS channel, the IRS reflection matrix, and the IRS-user channel. Both direct and IRS-aided link contributes to the composite channel $\mathbf{H} \in \mathbb{C}^{K \times N}$ as

$$\mathbf{H} = \mathbf{H}_D + \mathbf{H}_R \mathbf{\Theta} \mathbf{H}_I \quad (7)$$

whose (k, n) -th entry

$$h_{k,n} = h_{D,k,n} + \mathbf{h}_{R,k,n} \mathbf{\Theta}_0 \mathbf{h}_{I,n} \quad (8)$$

represents the overall channel gain of user k at subband n . On top of this, the n -th subchannel for all users $\mathbf{h}_n \in \mathbb{C}^{K \times 1}$ can be expressed as

$$\mathbf{h}_n = \mathbf{h}_{D,n} + \mathbf{H}_{R,n} \mathbf{\Theta}_0 \mathbf{h}_{I,n} \quad (9)$$

C. Received Signal

The RF signal received by user k captures the contribution of information and power waveforms through both direct and reflective links as

$$y_k(t) = \sum_{j=1}^K \sum_{n=1}^N h_{k,n} (w_{I,j,n} \tilde{x}_{I,j,n} + w_{P,j,n}) e^{j2\pi f_n t} \quad (10)$$

with the information and power components of

$$y_{I,k}(t) = \sum_{j=1}^K \sum_{n=1}^N h_{k,n} w_{I,j,n} \tilde{x}_{I,j,n} e^{j2\pi f_n t} \quad (11)$$

$$y_{P,k}(t) = \sum_{j=1}^K \sum_{n=1}^N h_{k,n} w_{P,j,n} e^{j2\pi f_n t} \quad (12)$$

D. Information Decoder

One major benefit of using proposed waveform is that the deterministic power term $y_{P,k}(t)$ bears no information and creates no interference to the modulated information term $y_{I,k}(t)$. Hence, each user treats the information signals from other users as interference such that the SINR at subband n of user k is

$$\gamma_{k,n} = \frac{(1 - \rho)(h_{k,n} w_{I,k,n})(h_{k,n} w_{I,k,n})^*}{\sum_{j=1, j \neq k}^K (1 - \rho)(h_{k,n} w_{I,j,n})(h_{k,n} w_{I,j,n})^* + \sigma_n^2} \quad (13)$$

where σ_n^2 is the sum variance of the Gaussian noise at the RF-band and those introduced during the RF-to-BB conversion on subband n . Therefore, the achievable rate of user k is expressed as

$$R_k(\mathbf{W}_I, \mathbf{\Theta}_0, \rho) = \sum_{n=1}^N \log_2(1 + \gamma_{k,n}) \quad (14)$$

A significant conclusion in [2] indicates that the rate 14 is always achievable with or without waveform cancellation, through either subtracting the deterministic power component or constructing a translated codebook.

E. Energy Harvester

Note that the information and power component $y_{I,k}(t)$ and $y_{P,k}(t)$ have different influence on the output DC current. Consider a nonlinear diode model based on the Taylor expansion of a small signal model [2], [7], which highlights the dependency of the harvester output DC current on the received waveform as

$$i_k \approx \sum_{i=0}^{\infty} k'_i \rho^{i/2} R_{\text{ant}}^{i/2} \mathcal{E}\{\mathcal{A}\{y_k(t)^i\}\} \quad (15)$$

where R_{ant} is the impedance of the receive antenna, $k'_0 = i_s(e^{-i_k R_{\text{ant}}/n v_t} - 1)$, $k'_i = i_s e^{-i_k R_{\text{ant}}/n v_t} / i! (n v_t)^i$ for $i = 1, \dots, \infty$, i_s is saturation current, n is diode ideality factor, v_t is thermal voltage. Once the composite channel response $h_{k,n}$ is fixed, the corresponding information and power weight $w_{I,k,n}$, $w_{P,k,n}$ are thus optimized such that the randomness comes from the input distribution $\tilde{x}_{k,n}$. Therefore, we first extract the DC component based on $h_{k,n}$, $w_{I,k,n}$, $w_{P,k,n}$ by $\mathcal{A}\{\cdot\}$ and then take the expectation over $\tilde{x}_{k,n}$ by $\mathcal{E}\{\cdot\}$. With the assumption of evenly spaced frequencies, $\mathcal{E}\{y^i(t)\} = 0$ for odd i and the related terms has zero contribution to DC components. [7] also demonstrated that to maximize i_k , it suffices to maximize the monotonic target function truncated to the n_0 order

$$z_k(\mathbf{W}_I, \mathbf{W}_P, \mathbf{\Theta}_0, \rho) = \sum_{i \text{ even}, i \geq 2}^{n_0} k_i \rho^{i/2} R_{\text{ant}}^{i/2} \mathcal{E}\{\mathcal{A}\{y_k(t)^i\}\} \quad (16)$$

where $k_i = i_s / i! (n v_t)^i$. We take $n_0 = 4$ to investigate the fundamental impact of diode nonlinearity on power transfer. Note that $\mathcal{E}\{\mathcal{A}\{y_{I,k}(t) y_{P,k}(t)\}\} = \mathcal{E}\{\mathcal{A}\{y_{I,k}^3(t) y_{P,k}(t)\}\} = \mathcal{E}\{\mathcal{A}\{y_{I,k}(t) y_{P,k}^3(t)\}\} = 0$, $\mathcal{E}\{\mathcal{A}\{y_{I,k}^2(t) y_{P,k}^2(t)\}\} = \mathcal{A}\{\mathcal{E}\{y_{I,k}^2(t)\} \mathcal{A}\{y_{P,k}^2(t)\}\}$, and the remaining terms are expressed by 17 – 24. It is noteworthy that modulation provides a power gain to the nonlinear terms in the output DC current as $\mathcal{E}\{|\tilde{x}_{k,n}|^2\} = 1$ and $\mathcal{E}\{|\tilde{x}_{k,n}|^4\} = 2$. Therefore, 16 reduces to 25.

II. PROBLEM FORMULATION

In this section, we first define the weighted sum rate-energy (WSR-E) region then formulate waveform and IRS optimization problems.

A. Weighted Sum Rate-Energy Region

We define the achievable WSR-E region as

$$C_{R-I}(P) \triangleq \left\{ (R, I) : R \leq \sum_{k=1}^K u_{I,k} R_k, I \leq \sum_{k=1}^K u_{P,k} z_k, \right. \\ \left. \frac{1}{2}(\mathbf{w}_I^H \mathbf{w}_I + \mathbf{w}_P^H \mathbf{w}_P) \leq P \right\} \quad (26)$$

where $u_{I,k}, u_{P,k}$ denote the information and power weight of user k .

B. Single-User Optimization

In this section, we first characterize the rate-energy region in the single-user setup. Define $h_n = A_n e^{j\psi_n}$, $w_{I,n} = s_{I,n} e^{j\phi_n}$, $w_{P,n} = s_{P,n} e^{j\phi_n}$, $\mathbf{v} = [e^{j\theta_1}, \dots, e^{j\theta_L}]^H \in \mathbb{C}^{L \times 1}$, and $\Phi_n = \text{diag}\{\mathbf{h}_{R,n}\} \mathbf{h}_{I,n} \in \mathbb{C}^{L \times 1}$. On top of this, the composite channel \mathbf{h} can be expressed as $h_n = h_{D,n} + \mathbf{v}^H \Phi_n$. The WSR-E region can be obtained through a rate maximization

problem subject to transmit power and output DC current constraints

$$\begin{aligned} \max_{\mathbf{w}_I, \mathbf{w}_P, \mathbf{v}, \rho} \quad & \sum_n \log_2 \left(1 + \frac{(1-\rho)(h_{D,n} + \mathbf{v}^H \Phi_n) w_{I,n}}{\sigma_n^2} \right) \\ \text{s.t.} \quad & \frac{1}{2}(\mathbf{w}_I^H \mathbf{w}_I + \mathbf{w}_P^H \mathbf{w}_P) \leq P, \\ & z(\mathbf{w}_I, \mathbf{w}_P, \mathbf{v}, \rho) \geq \bar{z}, \\ & |v_l| = 1, \quad l = 1, \dots, L \end{aligned} \quad (27)$$

27 is nontrivial due to the non-convex objective function, unit-modulus constraint, and coupled variables. To reduce the design complexity, we propose an suboptimal alternating optimization algorithm that updates the phase shifts at the IRS and the information weight at the transmitter iteratively until convergence.

The phase optimization problem can be formed as follows. For a fixed information weight \mathbf{w}_I and power-splitting ratio

$$\mathcal{E}\{\mathcal{A}\{y_{I,k}^2(t)\}\} = \frac{1}{2} \sum_{j_1, j_2} \sum_n (h_{k,n} w_{I,j_1,n})(h_{k,n} w_{I,j_2,n})^* \quad (17)$$

$$= \frac{1}{2} \sum_n h_{k,n} h_{k,n}^* \mathbf{w}_{I,n}^H \mathbf{J}_K \mathbf{w}_{I,n} \quad (18)$$

$$\mathcal{E}\{\mathcal{A}\{y_{I,k}^4(t)\}\} = \frac{3}{4} \sum_{j_1, j_2, j_3, j_4} \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} (h_{k,n_1} w_{I,j_1,n_1})(h_{k,n_2} w_{I,j_2,n_2})(h_{k,n_3} w_{I,j_3,n_3})^* (h_{k,n_4} w_{I,j_4,n_4})^* \quad (19)$$

$$= \frac{3}{4} \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} h_{k,n_3}^* h_{k,n_1} h_{k,n_4}^* h_{k,n_2} (\mathbf{w}_{I,n_3}^H \mathbf{J}_K \mathbf{w}_{I,n_1})(\mathbf{w}_{I,n_4}^H \mathbf{J}_K \mathbf{w}_{I,n_2}) \quad (20)$$

$$\mathcal{A}\{y_{P,k}^2(t)\} = \frac{1}{2} \sum_{j_1, j_2} \sum_n (h_{k,n} w_{P,j_1,n})(h_{k,n} w_{P,j_2,n})^* \quad (21)$$

$$= \frac{1}{2} \sum_n h_{k,n} h_{k,n}^* \mathbf{w}_{P,n}^H \mathbf{J}_K \mathbf{w}_{P,n} \quad (22)$$

$$\mathcal{A}\{y_{P,k}^4(t)\} = \frac{3}{8} \sum_{j_1, j_2, j_3, j_4} \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} (h_{k,n_1} w_{P,j_1,n_1})(h_{k,n_2} w_{P,j_2,n_2})(h_{k,n_3} w_{P,j_3,n_3})^* (h_{k,n_4} w_{P,j_4,n_4})^* \quad (23)$$

$$= \frac{3}{8} \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} h_{k,n_3}^* h_{k,n_1} h_{k,n_4}^* h_{k,n_2} (\mathbf{w}_{P,n_3}^H \mathbf{J}_K \mathbf{w}_{P,n_1})(\mathbf{w}_{P,n_4}^H \mathbf{J}_K \mathbf{w}_{P,n_2}) \quad (24)$$

$$\begin{aligned} z_k &= k_2 \rho R_{\text{ant}} (\mathcal{E}\{\mathcal{A}\{y_{I,k}^2(t)\}\} + \mathcal{A}\{y_{P,k}^2(t)\}) + k_4 \rho^2 R_{\text{ant}}^2 (\mathcal{E}\{\mathcal{A}\{y_{I,k}^4(t)\}\} + \mathcal{A}\{y_{P,k}^4(t)\}) + 6 \mathcal{E}\{\mathcal{A}\{y_{I,k}^2(t)\}\} \mathcal{A}\{y_{P,k}^2(t)\}) \\ &= \frac{1}{2} k_2 \rho R_{\text{ant}} \sum_n h_{k,n} h_{k,n}^* (\mathbf{w}_{I,n}^H \mathbf{J}_K \mathbf{w}_{I,n} + \mathbf{w}_{P,n}^H \mathbf{J}_K \mathbf{w}_{P,n}) \\ &\quad + \frac{3}{8} k_4 \rho^2 R_{\text{ant}}^2 \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} h_{k,n_3}^* h_{k,n_1} h_{k,n_4}^* h_{k,n_2} (2(\mathbf{w}_{I,n_3}^H \mathbf{J}_K \mathbf{w}_{I,n_1})(\mathbf{w}_{I,n_4}^H \mathbf{J}_K \mathbf{w}_{I,n_2}) + (\mathbf{w}_{P,n_3}^H \mathbf{J}_K \mathbf{w}_{P,n_1})(\mathbf{w}_{P,n_4}^H \mathbf{J}_K \mathbf{w}_{P,n_2})) \\ &\quad + \frac{3}{2} k_4 \rho^2 R_{\text{ant}}^2 \sum_n h_{k,n}^2 h_{k,n}^2 (\mathbf{w}_{I,n}^H \mathbf{J}_K \mathbf{w}_{I,n})(\mathbf{w}_{P,n}^H \mathbf{J}_K \mathbf{w}_{P,n}) \end{aligned} \quad (25)$$

ρ , problem 27 reduces to

$$\max_{\mathbf{v}} \sum_n \log_2 \left(1 + \frac{(1-\rho)|(h_{D,n} + \mathbf{v}^H \Phi_n)|^2 |w_{I,n}|^2}{\sigma_n^2} \right) \quad (28)$$

Therefore, 27 is transformed to

$$\begin{aligned} \max_{\mathbf{s}_I, \mathbf{s}_P, \rho} \quad & \sum_n \log_2 \left(1 + \frac{(1-\rho)A_n^2 s_{I,n}^2}{\sigma_n^2} \right) \\ \text{s.t.} \quad & \frac{1}{2}(\mathbf{s}_I^H \mathbf{s}_I + \mathbf{s}_P^H \mathbf{s}_P) \leq P, \\ & z(\mathbf{s}_I, \mathbf{s}_P, \rho) \geq \bar{z} \end{aligned} \quad (29)$$

with z given by 30. It can be transformed to an equivalent problem by introducing an auxiliary variable t_0

$$\begin{aligned} \min_{\mathbf{s}_I, \mathbf{s}_P, \rho} \quad & \frac{1}{t_0} \\ \text{s.t.} \quad & \frac{1}{2}(\mathbf{s}_I^H \mathbf{s}_I + \mathbf{s}_P^H \mathbf{s}_P) \leq P, \\ & \frac{t_0}{\prod_n \left(1 + \frac{(1-\rho)A_n^2 s_{I,n}^2}{\sigma_n^2} \right)} \leq 1, \\ & \frac{\bar{z}}{z(\mathbf{s}_I, \mathbf{s}_P, \rho)} \leq 1 \end{aligned} \quad (31)$$

31 is a Reversed Geometric Program which can be transformed to standard Geometric Program (GP). The basic idea is to decompose the information and power posynomials as sum of monomials, then derive their upper bounds using Arithmetic Mean-Geometric Mean (AM-GM) inequality [2], [8]. Let $z(\mathbf{s}_I, \mathbf{s}_P, \rho) = \sum_{m_P=1}^{M_P} g_{P,m_P}(\mathbf{s}_I, \mathbf{s}_P, \rho)$, problem 31 is equivalent to

$$\begin{aligned} \min_{\mathbf{s}_I, \mathbf{s}_P, \rho} \quad & \frac{1}{t_0} \\ \text{s.t.} \quad & \frac{1}{2}(\mathbf{s}_I^H \mathbf{s}_I + \mathbf{s}_P^H \mathbf{s}_P) \leq P, \\ & t_0 \prod_n \left(\frac{1}{\gamma_{I,n,1}} \right)^{-\gamma_{I,n,1}} \left(\frac{\bar{\rho} A_n^2 s_{I,n}^2}{\sigma_n^2 \gamma_{I,n,2}} \right)^{-\gamma_{I,n,2}} \leq 1, \\ & \bar{z} \prod_{m_P} \left(\frac{g_{P,m_P}(\mathbf{s}_I, \mathbf{s}_P, \rho)}{\gamma_{P,m_P}} \right)^{-\gamma_{P,m_P}} \leq 1, \\ & \rho + \bar{\rho} \leq 1 \end{aligned} \quad (32)$$

where $\gamma_{I,n,1}, \gamma_{I,n,2} \geq 0$, $\gamma_{I,n,1} + \gamma_{I,n,2} = 1$, $\gamma_{P,m_P} \geq 0, \forall m_P = 1, \dots, M_P$ and $\sum_{m_P=1}^{M_P} \gamma_{m_P} = 1$. As suggested in [2], the tightness of the AM-GM inequality depends on

$\{\gamma_{I,n}, \gamma_P\}$ that require iterative update. At iteration i , we choose

$$\gamma_{I,n,1}^{(i)} = \frac{1}{1 + \frac{\bar{\rho}^{(i-1)} A_n^2 s_{I,n}^{(i-1)2}}{\sigma_n^2}}, \quad n = 1, \dots, N, \quad (33)$$

$$\gamma_{I,n,2}^{(i)} = \frac{\frac{\bar{\rho}^{(i-1)} A_n^2 s_{I,n}^{(i-1)2}}{\sigma_n^2}}{1 + \frac{\bar{\rho}^{(i-1)} A_n^2 s_{I,n}^{(i-1)2}}{\sigma_n^2}}, \quad n = 1, \dots, N, \quad (34)$$

$$\gamma_{P,m_P}^{(i)} = \frac{g_{P,m_P}(\mathbf{s}_I^{(i-1)}, \mathbf{s}_P^{(i-1)}, \rho^{(i-1)})}{z(\mathbf{s}_I^{(i-1)}, \mathbf{s}_P^{(i-1)}, \rho^{(i-1)})}, \quad m_P = 1, \dots, M_P \quad (35)$$

and then solve problem 32.

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$$\begin{aligned} z = & \frac{1}{2} k_2 \rho R_{\text{ant}} \sum_n A_n^2 (s_{I,n}^2 + s_{P,n}^2) + \frac{3}{8} k_4 \rho^2 R_{\text{ant}}^2 \sum_{\substack{n_1, n_2, n_3, n_4 \\ n_1 + n_2 = n_3 + n_4}} A_{n_1} A_{n_2} A_{n_3} A_{n_4} (2s_{I,n_1} s_{I,n_2} s_{I,n_3} s_{I,n_4} + s_{P,n_1} s_{P,n_2} s_{P,n_3} s_{P,n_4}) \\ & + \frac{3}{2} k_4 \rho^2 R_{\text{ant}}^2 \sum_n A_n^4 s_{I,n}^2 s_{P,n}^2 \end{aligned} \quad (30)$$