Expanding the Compute-and-Forward Framework

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Compute-and-Forward: Single Receiver



- Messages are finite field vectors, $\mathbf{w}_{\ell} \in \mathbb{Z}_p^k$, p prime.
- Real-valued inputs and outputs, $\mathbf{x}_{\ell}, \mathbf{y} \in \mathbb{R}^{n}.$
- Equal power constraint, $\mathbb{E} \|\mathbf{x}_{\ell}\|^2 \leq nP.$
- Gaussian noise, $\mathbf{z} \sim \mathcal{N}(\mathbf{0}, \mathbf{I}).$

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- What rates are achievable as a function of h_{ℓ} and q_{ℓ} ?

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- The linear combination with integer coefficient vector $\mathbf{a} = [a_1 \ a_2 \ \cdots \ a_L]^T \in \mathbb{Z}^L$ corresponds to $\mathbf{u} = \bigoplus_{\ell=1}^L q_\ell \mathbf{w}_\ell$ where $q_\ell = [a_\ell] \mod p$.

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- Key Definition: The computation rate region described by $R_{\text{comp}}(\mathbf{h}, \mathbf{a})$ is achievable if, for any $\epsilon > 0$ and n, p large enough, the receiver can decode any linear combination with integer coefficient vector $\mathbf{a} \in \mathbb{Z}^L$ with probability of error at most ϵ so long as the message rate R satisfies

$$R < R_{\mathsf{comp}}(\mathbf{h}, \mathbf{a})$$

Compute-and-Forward: Effective Noise

$$\mathbf{y} = \sum_{\ell=1}^{L} h_{\ell} \mathbf{x}_{\ell} + \mathbf{z}$$
$$= \sum_{\ell=1}^{L} a_{\ell} \mathbf{x}_{\ell} + \sum_{\ell=1}^{L} (h_{\ell} - a_{\ell}) \mathbf{x}_{\ell} + \mathbf{z}$$

Desired Codebook:

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- Independent effective noise \implies dithering.
- Isomorphic to $\mathbb{Z}_p^k \implies$ nested lattice codebook.

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- Quantization error serves as modulo operation:

$$[\mathbf{x}] \mod \Lambda = \mathbf{x} - Q_{\Lambda}(\mathbf{x}) .$$



Distributive Law:

 $\begin{bmatrix} \mathbf{x}_1 + a[\mathbf{x}_2] \mod \Lambda \end{bmatrix} \mod \Lambda = [\mathbf{x}_1 + a\mathbf{x}_2] \mod \Lambda \quad \text{for all } a \in \mathbb{Z}.$

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- Existence of good nested lattice codes: Loeliger '97, Forney-Trott-Chung '00, Erez-Litsyn-Zamir '05, Ordentlich-Erez '12.
- Erez-Zamir '04: Nested lattice codes can achieve the Gaussian capacity.
- Zamir-Shamai-Erez '02: Excellent framework for multi-terminal binning.



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- Decode integer-linear combination $\left[\sum_{\ell} a_{\ell} \lambda_{\ell}\right] \mod \Lambda_{c}$ from

$$\beta \mathbf{y} = \sum_{\ell=1}^{L} a_{\ell} \mathbf{x}_{\ell} + \underbrace{\sum_{\ell=1}^{L} (\beta h_{\ell} - a_{\ell}) \mathbf{x}_{\ell} + \beta \mathbf{z}}_{\text{effective noise}} \,.$$

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• Succeeds w.h.p. if noise tolerance σ_{eff}^2 of the fine lattice satisfies

$$\sigma_{\mathsf{eff}}^2 > \min_{\beta \in \mathbb{R}} \left(\beta^2 + P \|\beta \mathbf{h} - \mathbf{a}\|^2 \right) = \left\| \left(P^{-1} \mathbf{I} + \mathbf{h} \mathbf{h}^\mathsf{T} \right)^{-1/2} \mathbf{a} \right\|^2 \,.$$

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• Map $\left[\sum_{\ell} a_{\ell} \lambda_{\ell}\right] \mod \Lambda_{c}$ back to \mathbb{Z}_{p} to get $\mathbf{u} = \bigoplus_{\ell} q_{\ell} \mathbf{w}_{\ell}$.

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- Map $\left[\sum_{\ell} a_{\ell} \lambda_{\ell}\right] \mod \Lambda_{c}$ back to \mathbb{Z}_{p} to get $\mathbf{u} = \bigoplus_{\ell} q_{\ell} \mathbf{w}_{\ell}$.
- Nazer-Gastpar IT '11: The achievable computation rate is

$$R_{\mathsf{comp}}(\mathbf{h}, \mathbf{a}) = \frac{1}{2} \log^+ \left(\frac{P}{\left\| \left(P^{-1} \mathbf{I} + \mathbf{h} \mathbf{h}^{\mathsf{T}} \right)^{-1/2} \mathbf{a} \right\|^2} \right)$$

Compute-and-Forward: Multiple Receivers



- Equal power constraints and Gaussian noise as before.
- Since some receivers will see better channels than others, it will be useful to allow for different rates R₁,..., R_L. How can we retain the connection to Z_p?

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- Specifically, the message w_ℓ of the ℓth encoder consists of k_ℓ information symbols over Z_p followed by k k_ℓ zeros where k ≜ max_ℓ k_ℓ. The message rate is R_ℓ = k_ℓ/n log p.

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- The computation rate region described by $R_{\text{comp}}(\mathbf{h}, \mathbf{a})$ is achievable if, for any $\epsilon > 0$ and n, p large enough, each receiver can decode any linear combination with integer coefficient vector $\mathbf{a}^{[1]}, \ldots, \mathbf{a}^{[K]} \in \mathbb{Z}^L$ with probability of error at most ϵ so long as

$$R_\ell < \min_{i:a_\ell^{[i]}
eq 0} R_{\mathsf{comp}}(\mathbf{h}^{[i]}, \mathbf{a}^{[i]})$$

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- Furthermore, can show that if the last $k k_{\ell}$ elements of a vector in \mathbb{Z}_p^k are zero, then it will be mapped to $[\Lambda_{\mathsf{F},\ell}/\Lambda_{\mathsf{C}}]$.

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- Furthermore, can show that if the last $k k_{\ell}$ elements of a vector in \mathbb{Z}_p^k are zero, then it will be mapped to $[\Lambda_{\mathsf{F},\ell}/\Lambda_{\mathsf{C}}]$.
- Nazer-Gastpar IT '11: Overall, we can combine these codebooks with the techniques for the single receiver case to get that the following computation rate region is achievable:

$$R_{\mathsf{comp}}(\mathbf{h}, \mathbf{a}) = \frac{1}{2} \log^+ \left(\frac{P}{\left\| \left(P^{-1} \mathbf{I} + \mathbf{h} \mathbf{h}^{\mathsf{T}} \right)^{-1/2} \mathbf{a} \right\|^2} \right)$$

What's missing?

• Unequal power constraints:

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- Can we use this for the general compute-and-forward problem? While retaining the connection to the finite field?



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- At each transmitter, use the same fine lattice and a different coarse lattice, chosen to meet the power constraint.
- Can we use this for the general compute-and-forward problem? While retaining the connection to the finite field?
- Zhu-Gastpar IZS '14: Proposed a way to use this technique for compute-and-forward without a connection to the finite field.

What's missing?



 Decoding multiple linear combinations at a single receiver is a useful technique for MIMO decoding (Zhan-Nazer-Erez-Gastpar IT '14) and interference alignment (Ordentlich-Erez-Nazer IT '14, Ntranos-Cadambe-Nazer-Caire ISIT '13).

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- Decoding multiple linear combinations at a single receiver is a useful technique for MIMO decoding (Zhan-Nazer-Erez-Gastpar IT '14) and interference alignment (Ordentlich-Erez-Nazer IT '14, Ntranos-Cadambe-Nazer-Caire ISIT '13).
- After a receiver has decoded one or more linear combinations, it can use these as side information to help decode the rest (Nazer IZS '12, Ordentlich-Erez-Nazer IT '14, Ordentlich-Erez-Nazer Allerton '13).

Expanding Compute-and-Forward: Single Receiver



- Include unequal power constraints $\mathbb{E} \|\mathbf{x}_{\ell}\|^2 \leq nP_{\ell}$ and multiple antennas at the receiver.
- Relax to linear combinations of cosets.
- WLOG receiver wants L linear combinations (since we can set coefficients to 0).

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- Transmitter sets top $k_{C,\ell} k_C$ symbols to zero to meet its power constraint.
- Transmitter sets bottom $k_{\mathsf{F}} k_{\mathsf{F},\ell}$ symbols to zero to meet its noise tolerance constraint.
- Remaining $k_{F,\ell} k_{C,\ell}$ symbols carry information. Rate is $R_{\ell} = \frac{k_{F,\ell} k_{C,\ell}}{n} \log p$.



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• Receiver attempts to recover linear combinations of cosets:

$$\mathbf{u}_{m} = \bigoplus_{\ell=1}^{L} q_{m,\ell} \tilde{\mathbf{w}}_{\ell}$$
$$\tilde{\mathbf{w}}_{\ell} \in \llbracket \mathbf{w}_{\ell} \rrbracket \triangleq \left\{ \mathbf{w} \in \mathbb{Z}_{p}^{k} : \mathbf{w} = \begin{bmatrix} \mathbf{r} \\ \mathbf{w}_{\ell} \\ \mathbf{0}_{k_{\mathsf{F}}-k_{\mathsf{F},\ell}} \end{bmatrix} \text{ for some } \mathbf{r} \in \mathbb{Z}_{p}^{k_{\mathsf{C},\ell}-k_{\mathsf{C}}} \right\}$$
$$q_{m,1} \bigoplus_{\bullet} q_{m,2} \bigoplus_{\bullet} q_{m,2} \bigoplus_{\bullet} \cdots \bigoplus_{\bullet} q_{m,L} \bigoplus_{\bullet} q_{m,L}$$

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$$q_{m,1} \bigoplus_{\mathbf{0}} \bigoplus_{\mathbf{0}} q_{m,2} \bigoplus_{\mathbf{0}} \bigoplus_{\mathbf{0}} \cdots \bigoplus_{\mathbf{0}} q_{m,L} \bigoplus_{\mathbf{0}} \bigoplus_{\mathbf{0}}$$

 As before, a linear combination with integer coefficient vector a_m is one that satisfies [a_{m,ℓ}] mod p = q_{m,ℓ}.

Computation Rate Region

• We will specify the computation rate region via a set-valued function $\mathcal{R}_{comp}(\mathbf{H}, \mathbf{A})$ that maps each channel matrix $\mathbf{H} \in \mathbb{R}^{N_r \times L}$ and integer coefficient matrix $\mathbf{A} \in \mathbb{Z}^{L \times L}$ to a subset of \mathbb{R}^L_+ .



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- The computation rate region described by $\mathcal{R}_{\text{comp}}(\mathbf{H}, \mathbf{A})$ is achievable if, for every rate tuple $(R_1, R_2, \ldots, R_L) \in \mathbb{R}^L_+$, $\epsilon > 0$, and n large enough, we can select encoders and a decoder such that,
 - for all channel matrices $\mathbf{H} \in \mathbb{R}^{N_r \times L}$ and
 - every coefficient matrix $\mathbf{Q} \in \mathbb{Z}_p^{L \times L}$ for which there exists an integer matrix \mathbf{A} satisfying $(R_1, R_2, \dots, R_L) \in \mathcal{R}_{\mathsf{comp}}(\mathbf{H}, \mathbf{A})$ and $[\mathbf{A}] \mod p = \mathbf{Q}$,

the probability of decoding error is at most ϵ .

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- Transmitter ℓ uses nested lattice code $\left[\Lambda_{\mathsf{F},\ell}/\Lambda_{\mathsf{C},\ell}\right]$.
- Linear labeling idea from Chen-Silva-Kschischang IT '13 allows us to build mapping between cosets and nested lattice codes.

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• Effective noise:

$$\begin{split} \sigma_{\mathsf{para}}^2(\mathbf{H}, \mathbf{a}_m) &= \min_{\mathbf{b}_m} \|\mathbf{b}_m\|^2 + \left\| \left(\mathbf{b}_m^\mathsf{T} \mathbf{H} - \mathbf{a}_m^\mathsf{T} \right) \mathbf{P}^{1/2} \right\|^2 \\ &= \left\| \left(\mathbf{P}^{-1} + \mathbf{H}^\mathsf{T} \mathbf{H} \right)^{-1/2} \mathbf{a}_m \right\|^2 \end{split}$$

Theorem (Nazer-Cadambe-Ntranos-Caire '15)

For an AWGN network with L transmitters, a receiver, and power constraints P_1, P_2, \ldots, P_L , the following computation rate region is achievable,

$$\begin{split} \mathcal{R}_{comp}^{(para)}(\mathbf{H},\mathbf{A}) &= \bigcup_{\tilde{\mathbf{A}} \in \mathbb{Z}^{L \times L} \\ \operatorname{rowspan}(\mathbf{A}) \subseteq \operatorname{rowspan}(\tilde{\mathbf{A}}) \\ \mathcal{R}_{para}(\mathbf{H},\tilde{\mathbf{A}}) &= \left\{ (R_1, \dots, R_L) \in \mathbb{R}_+^L : \\ R_{\ell} \leq \frac{1}{2} \log^+ \left(\frac{P_{\ell}}{\sigma_{para}^2(\mathbf{H},\tilde{\mathbf{a}}_m)} \right) \, \forall (m,\ell) \, \text{ s.t. } \tilde{a}_{m,\ell} \neq 0 \right\} \end{split}$$

• Can rebuild real sum $\mathbf{a}_m^\mathsf{T} \mathbf{X}$ from \mathbf{u}_m and \mathbf{Y} .

"Successive" Computation

- Can rebuild real sum $\mathbf{a}_m^\mathsf{T} \mathbf{X}$ from \mathbf{u}_m and \mathbf{Y} .
- This allows us to build a better effective channel output:

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Effective noise:

$$\sigma_{\mathsf{succ}}^{2}(\mathbf{H}, \mathbf{a}_{m} | \mathbf{A}_{m-1}) = \min_{\mathbf{b}_{m}, \mathbf{c}_{m}} \|\mathbf{b}_{m}\|^{2} + \left\| \left(\mathbf{b}_{m}^{\mathsf{T}} \mathbf{H} + \mathbf{c}_{m}^{\mathsf{T}} \mathbf{A}_{m-1} - \mathbf{a}_{m}^{\mathsf{T}}\right) \mathbf{P}^{1/2} \right\|^{2}$$
$$= \left\| \mathbf{N}_{m-1} \left(\mathbf{P}^{-1} + \mathbf{H}^{\mathsf{T}} \mathbf{H} \right)^{-1/2} \mathbf{a}_{m} \right\|^{2}$$

where N_{m-1} is the nullspace projection corresponding to A_{m-1} .

• Algebraic Successive Cancellation: We can use $\mathbf{u}_1, \ldots, \mathbf{u}_{m-1}$ to eliminate certain codewords from our observation, and thus remove the constraints on them.

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- Let *I* ⊂ {1,..., *L*} × {1,..., *L*} denote a set of index pairs. We say that *I* is an *admissible mapping* for **A** if there exists a real-valued, lower unitriangular matrix **L** ∈ ℝ^{L×L} such that the (*m*, *l*)th entry of **LA** is equal to zero for all (*m*, *l*) ∉ *I*.

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Multiple-Access via Computation



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• Under this framework, rate regions naturally generalize to multiple receivers:

$$\mathcal{R}_{\mathsf{comp}}^{(\mathsf{para})}(\mathbf{H}^{[1]}, \dots, \mathbf{H}^{[K]}, \mathbf{A}^{[1]}, \dots, \mathbf{A}^{[K]}) = \bigcap_{i=1}^{K} \mathcal{R}_{\mathsf{comp}}^{(\mathsf{para})}(\mathbf{H}^{[i]}, \mathbf{A}^{[i]})$$

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- He-Nazer-Shamai ISIT '14: Using this framework, we have found an uplink-downlink duality relationship for compute-and-forward. Allows us to build a connection to the work of Hong-Caire IT '13.
- Ntranos-Cadambe-Nazer-Caire ISIT '13: Used these ideas for integer-forcing interference alignment.
- Nazer-Gastpar ITW '14: Used the problem statement to bring compute-and-forward to the discrete memoryless setting.
- Can the algebraic perspective of Chen-Silva-Kschischang IT '13 be applied to the expanded problem?
- Currently trying to bring in more sophisticated multi-user techniques.