

## Communication-Efficient Collaborative Perception

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## Introduction & Background



> Autonomous system has lots of real-world applications.



autonomous vehicles



autonomous aerial vehicles

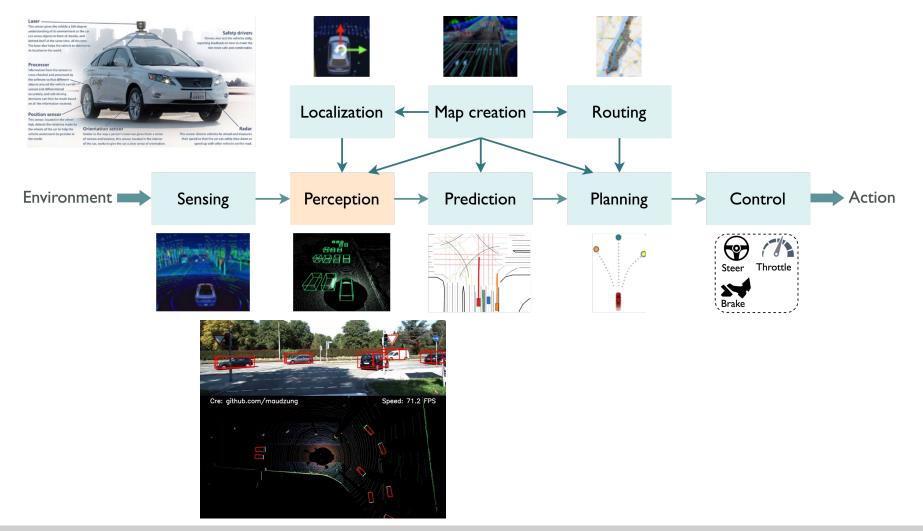


autonomous underwater vehicles



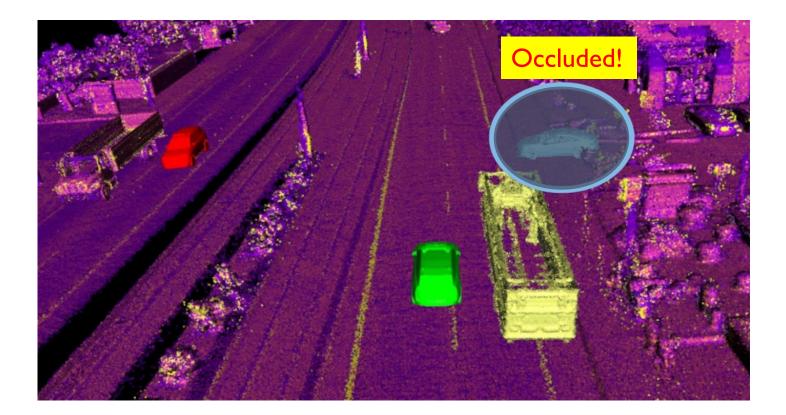
autonomous robotics

#### Perception is the front stage of the autonomous system.



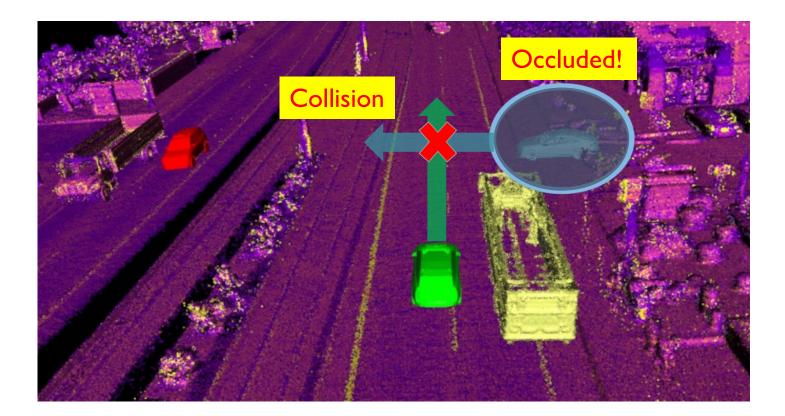
Single-agent perception

Single agent encounters limited visibility (occlusion) and long-range issues



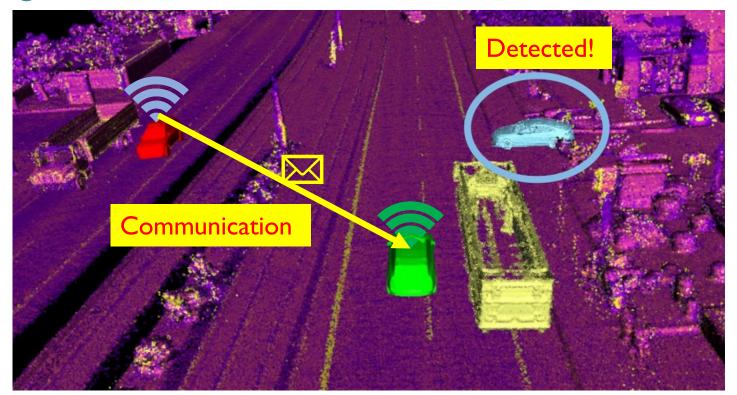
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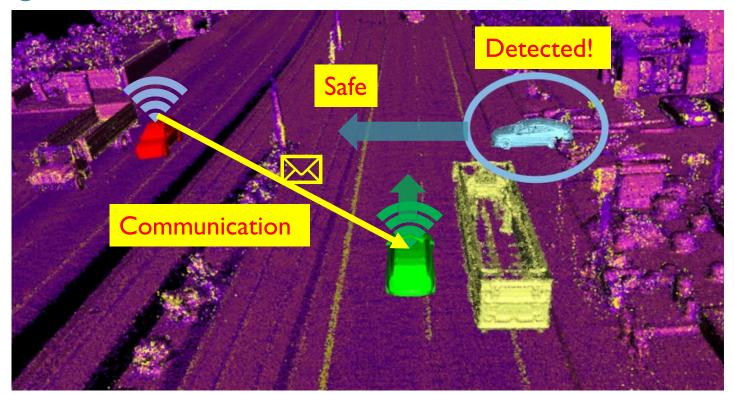
#### Multi-agent collaborative perception

**Collaborative agent fundamentally overcomes single-agent limitations,** achieving holistic view!

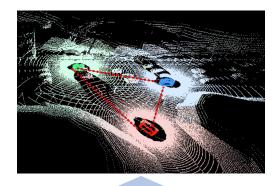


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Multi-agent collaborative perception



**Collaboration Perception** 



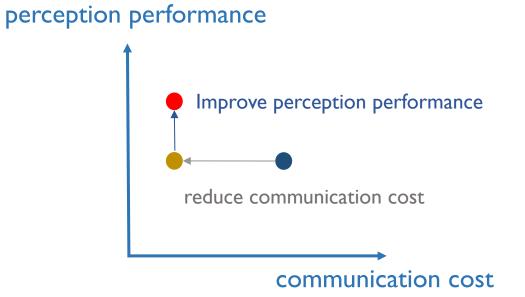
Single-Agent Perception Limited

Communication



Multi-agent collaborative perception

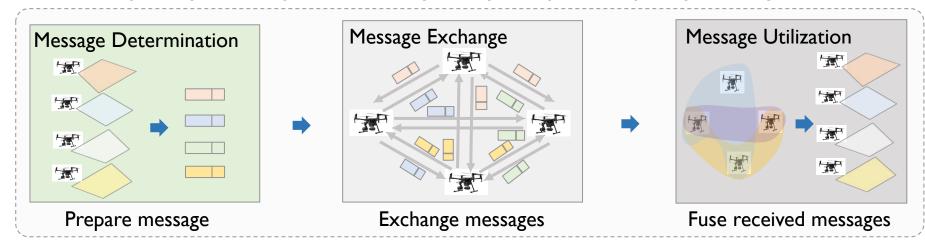
Key challenge: Trade-off between communication cost and perception performance

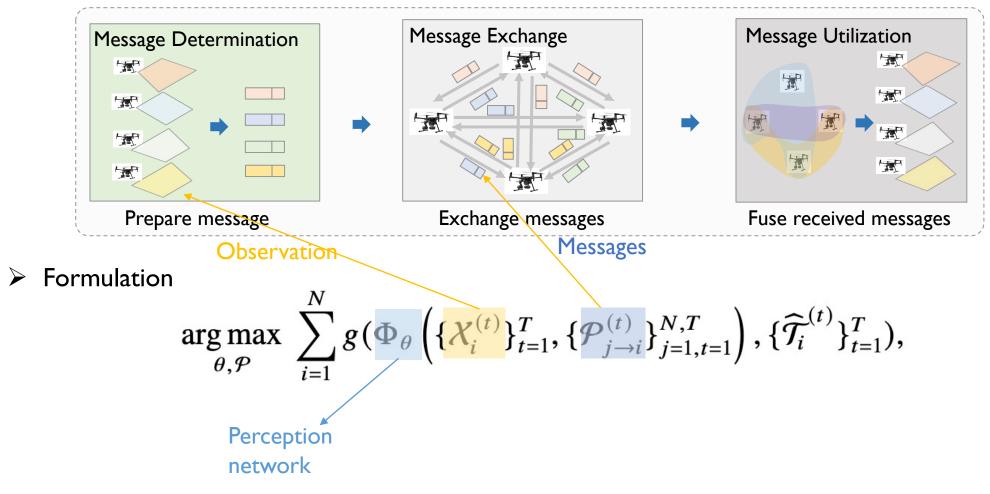


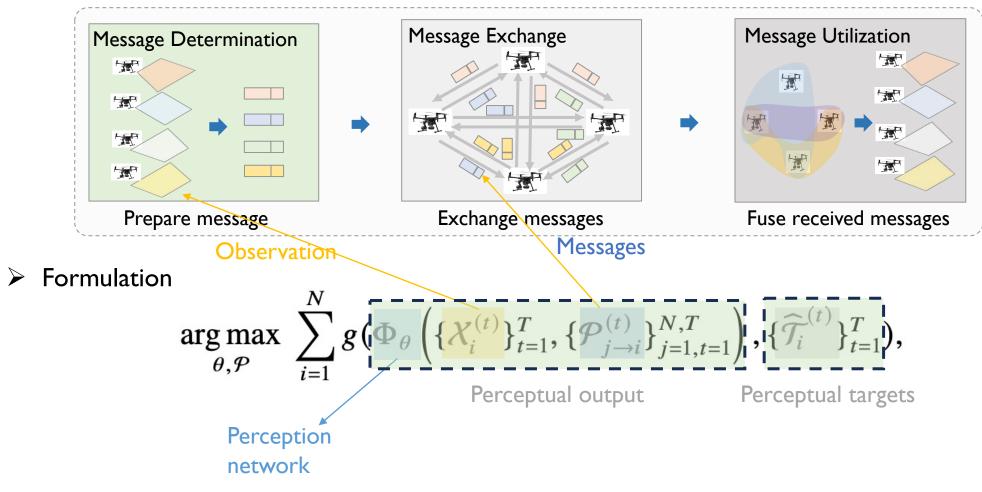
 $n^{2}$ 

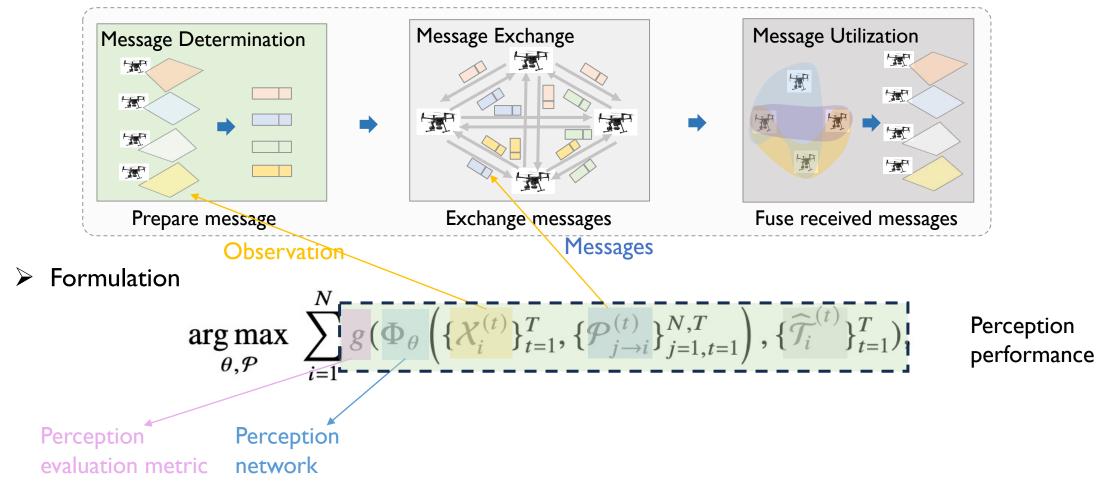
## **Collaborative Perception Formulation**



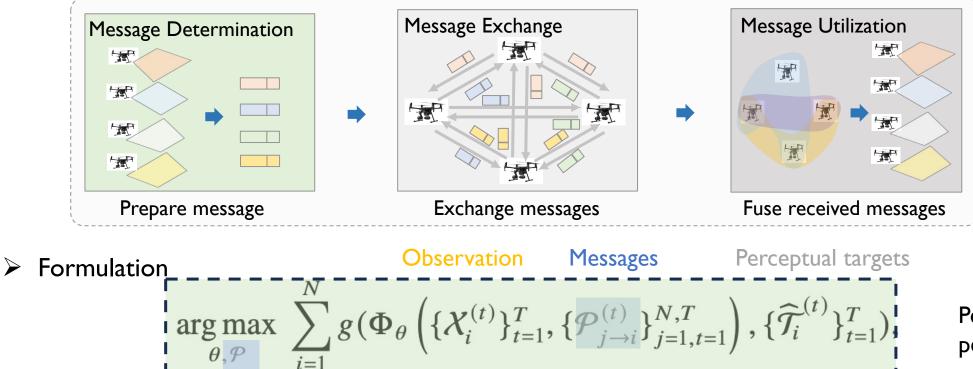






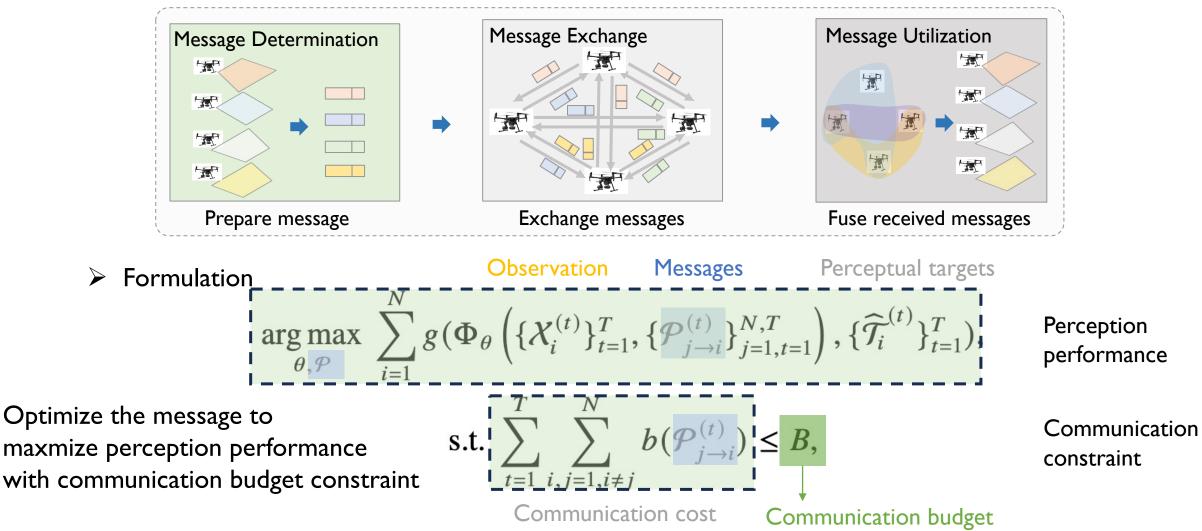


> Collaborative perception: improve each agent's perception capacity through information exchange

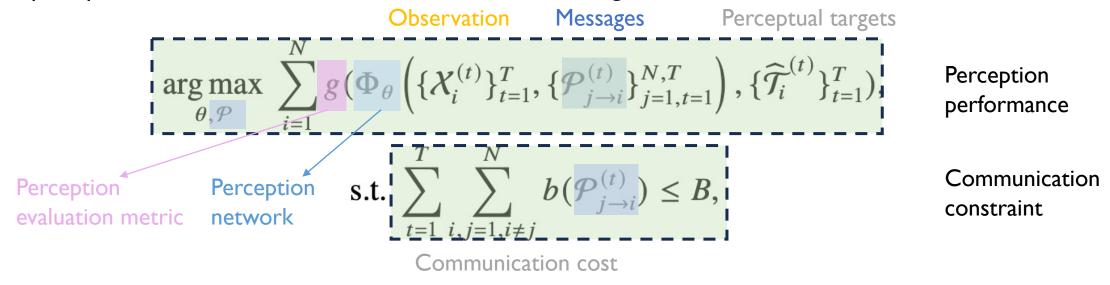


Perception performance

Optimize the message to maxmize perception performance



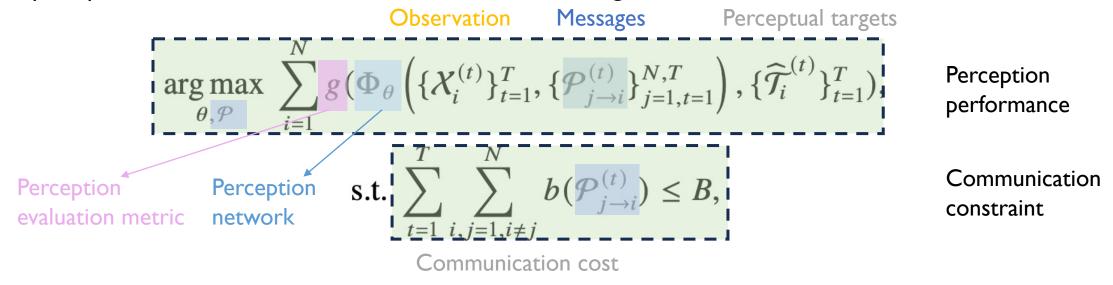
Task-specific communication objective: optimizing messages to fill each agent's specific perception task demand within the communication budget



#### Challenges:

- I) Non-differentiable perception metric and hard bandwidth constraint  $\rightarrow$  unable to directly optimize
- Diverse input and numerous perception network parameters → unable to enumerate the large searching space

Task-specific communication objective: optimizing messages to fill each agent's specific perception task demand within the communication budget



#### Solution: pragmatic communication

Key idea: acknowledge the pragmatic significance of the data and to develop pragmatic messages that retain only the data necessary for the collaborator's downstream task.

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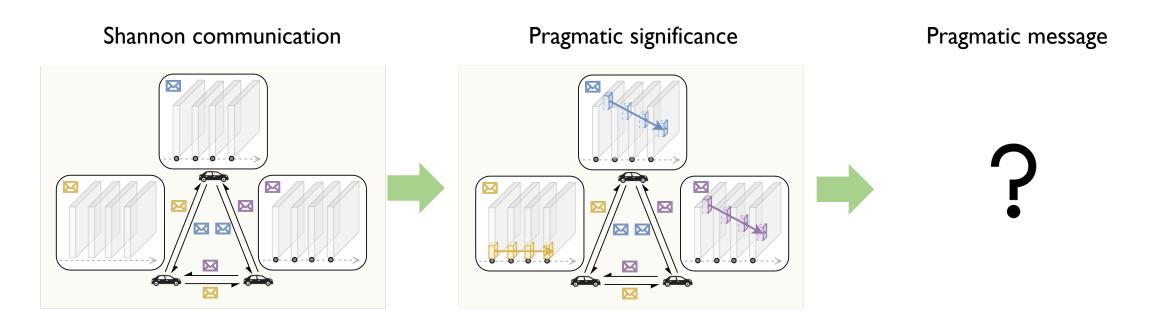
## Pragmatic Communication Solution



Where2comm: Communication-Efficient Collaborative Perception via Spatial Confidence Maps, NeurIPS, 2022
Pragmatic Communication in Multi-Agent Collaborative Perception, TPAMI, 2024 Submitted
Communication- Efficient Collaborative Perception via Information Filling with Codebook, CVPR, 2024

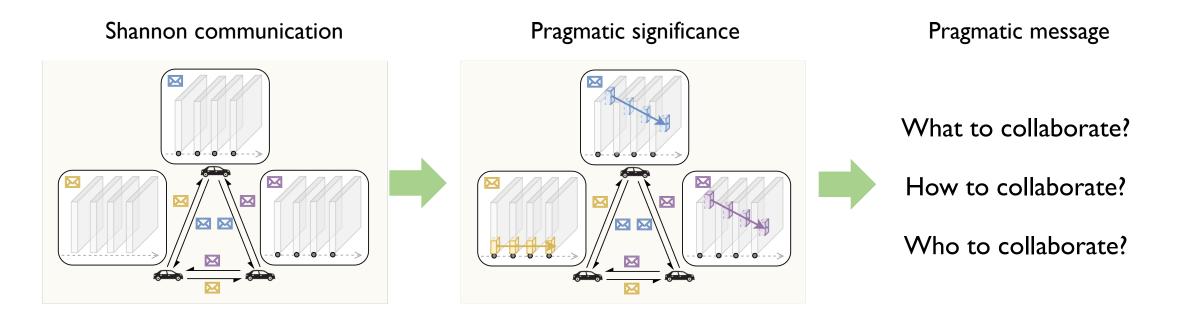
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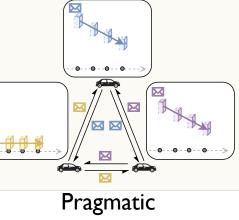


#### Pragmatic message

$$\mathcal{P} = \Psi_{\text{who}} \left( \Psi_{\text{represent}} \left( \Psi_{\text{select}} \left( \mathcal{X} \right) \right) \right)$$

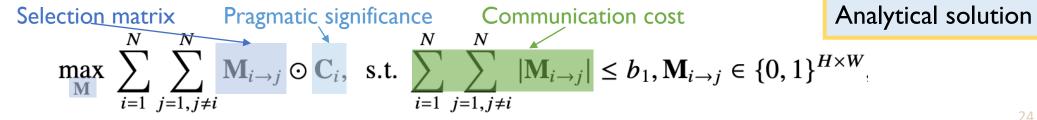
Shannon communication 0--0 . .  $\boxtimes$  $\bowtie$ 

What to collaborate?

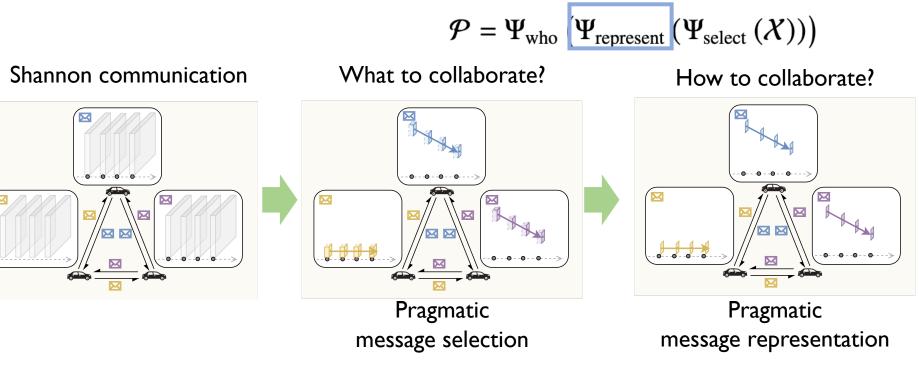


message selection

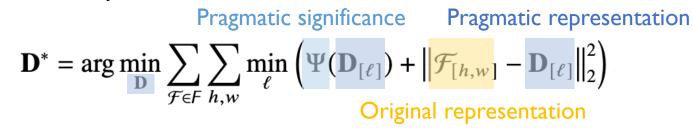
#### Reformulated objective



> Pragmatic message



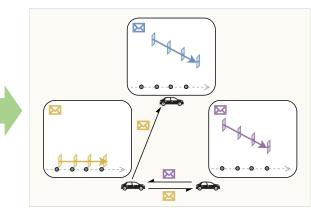
> Reformulated objective



Gradient decent

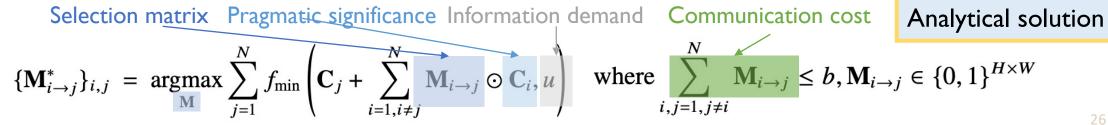
Pragmatic message  $\mathcal{P} = \Psi_{\text{who}} \left[ \Psi_{\text{represent}} \left( \Psi_{\text{select}} \left( \mathcal{X} \right) \right) \right]$ Shannon communication What to collaborate? How to collaborate? ----- $\bowtie$  $\mathbf{X}$  $\mathbf{X}$ - 🗪 Pragmatic Pragmatic message selection message representation

Who to collaborate?

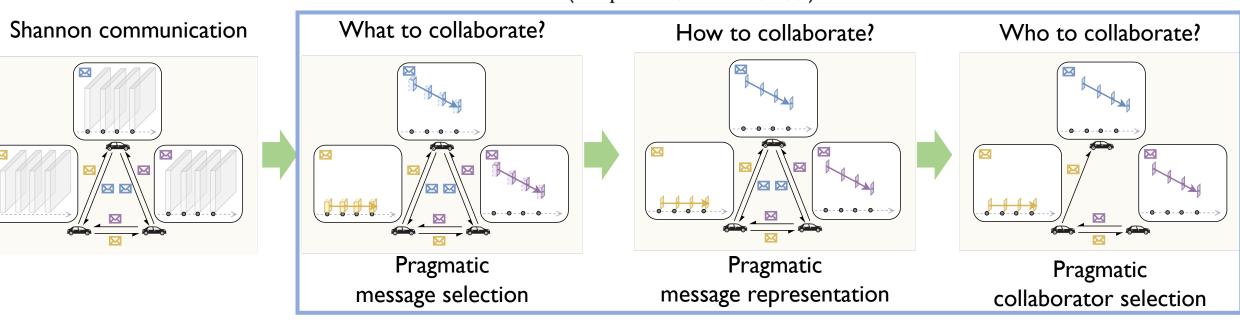


Pragmatic collaborator selection

#### **Reformulated objective** $\succ$



#### > Pragmatic message



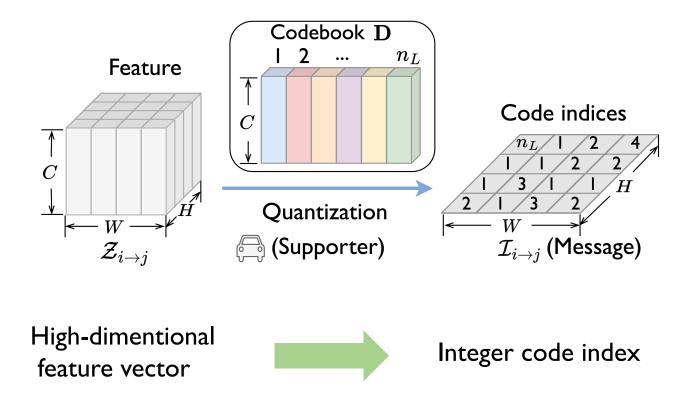
 $\mathcal{P} = \Psi_{\text{who}} \left( \Psi_{\text{represent}} \left( \Psi_{\text{select}} \left( \mathcal{X} \right) \right) \right)$ 

Pragmatic communication

Pragmatic messages retain only the data necessary for the collaborator's downstream task.

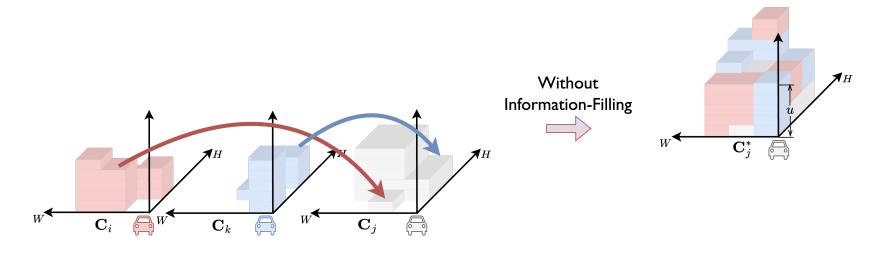
#### **Pragmatic Message Representation**

**Core idea:** leverage symbol-based communication (analogy to human language)



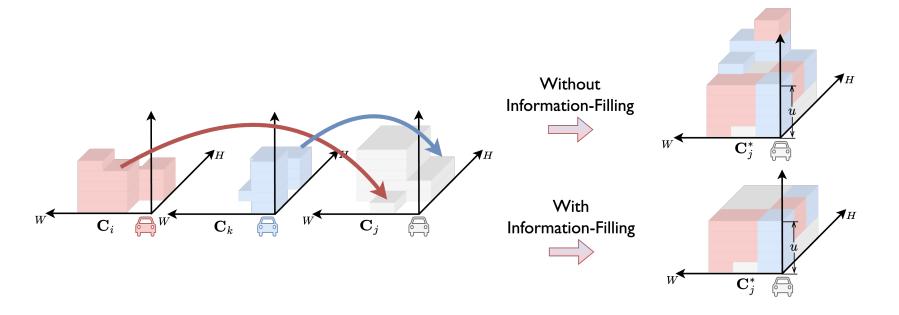
#### **Pragmatic Collaborator Selection**

**Core idea:** collectively fill each agent's information demand



#### **Pragmatic Collaborator Selection**

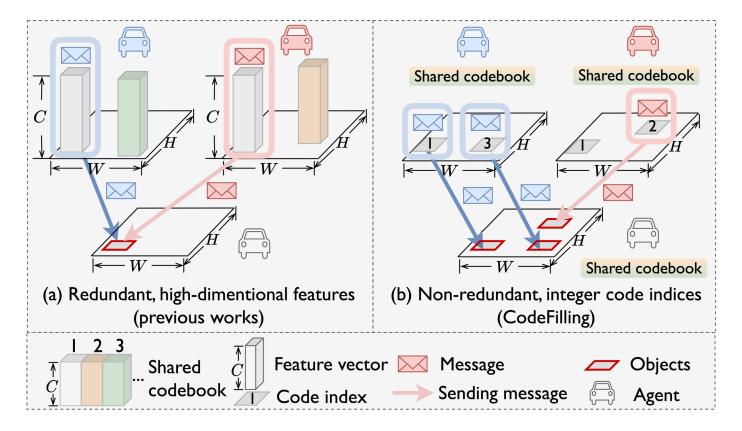
**Core idea:** collectively fill each agent's information demand



Avoid information overflow.

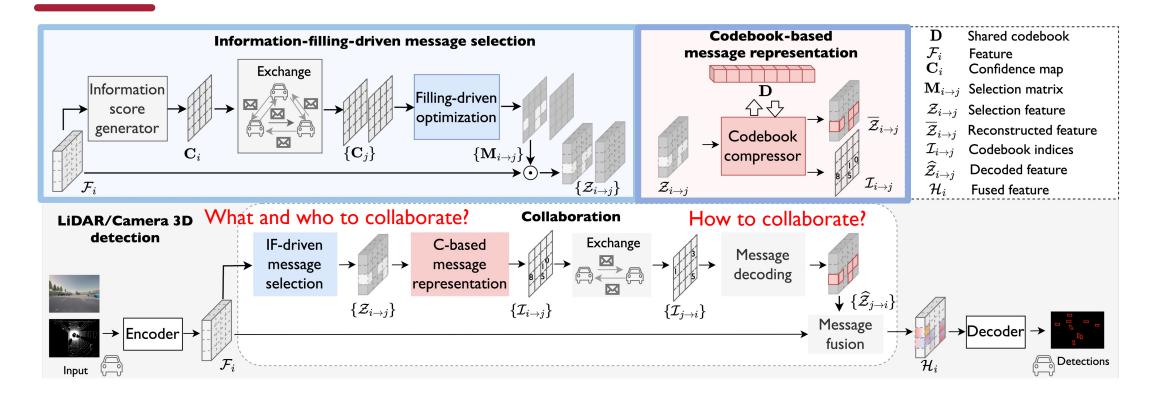
#### **Pragmatic Collaborator Selection**

**Core idea:** collectively fill each agent's information demand



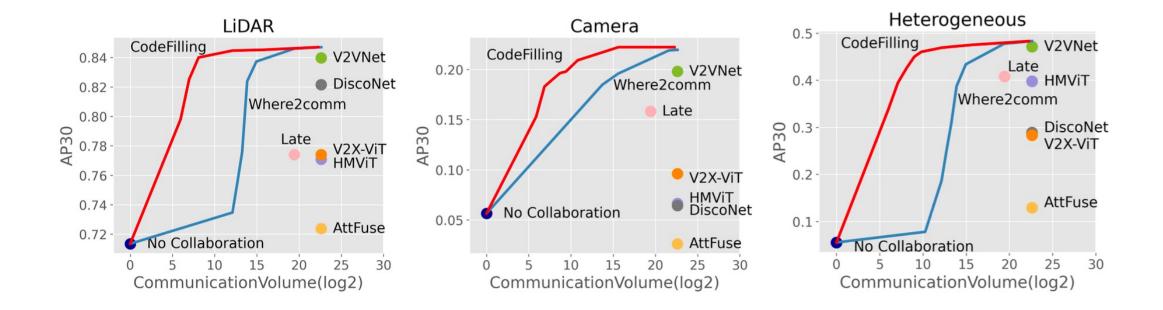
Reallocates the bandwidth wasted in redundant information to more beneficial information.

## **CodeFilling: Architecture**

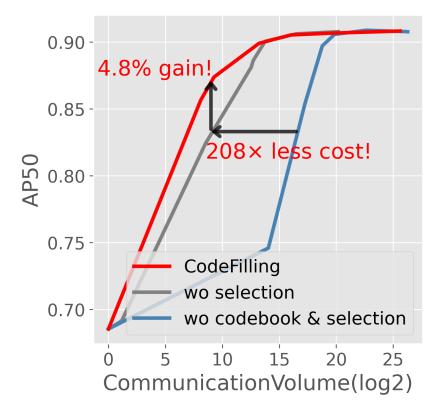


- Information-filling-driven message selection selects the **optimal collaborator set** at each spatial region to offer the most beneficial message.
- Codebook-based message representation represent the high-dimensional feature vector with integer code index.

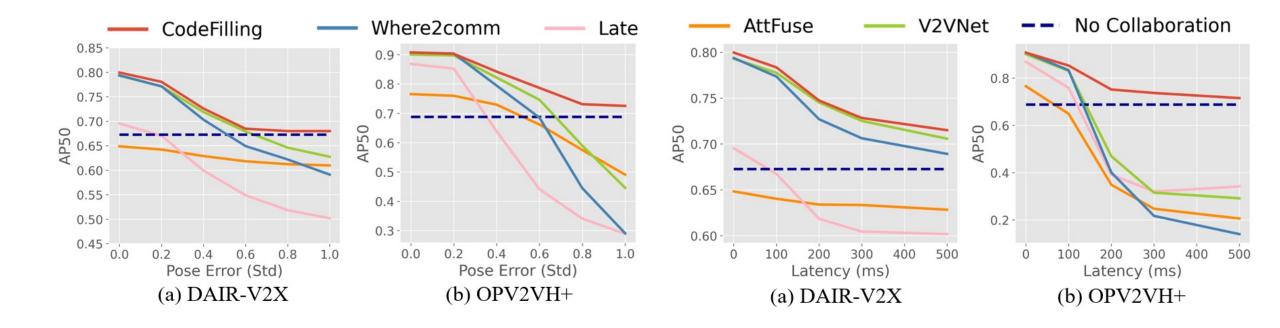
Compared to Where2comm, CodeFilling reduces communication cost under both homogeneous and heterogenous scenarios, and camera and LiDAR modality.



- > Compared to Where2comm, CodeFilling can further reduce the communication cost by 208 times.
- Pragmatic collaborator selection can improve the perception performance by 4.8% with the same communication cost.



> CodeFilling is robust to pose error and communication latency.



> Qualitative visualization of collaboration in CodeFilling.

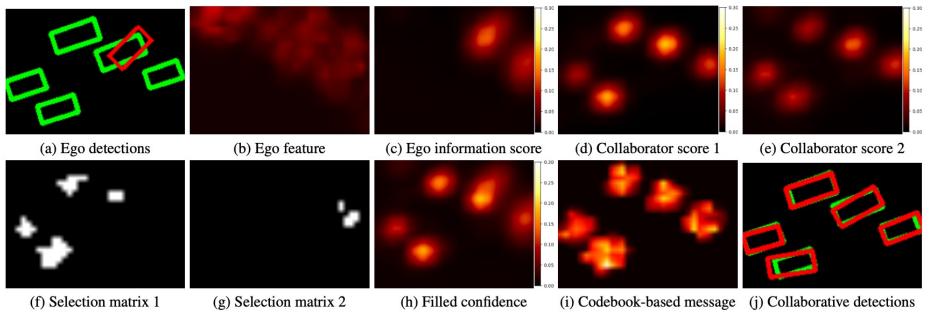


Figure 10. Visualization of collaboration in CodeFilling. Green and red denote ground truth and detection, respectively.

> CodeFilling qualitatively outperforms the state-of-the-art methods.



(a) No Colla

(b) V2X-ViT

(c) HMViT

(d) Where2comm

(e) CodeFilling