

Practical Optical Camera Communication Behind Unseen and Complex Backgrounds

Rui Xiao¹, Leqi Zhao¹, Feng Qian², Lei Yang², and Jinsong Han¹

¹ Zhejiang University

²Ant Group



Mixed Reality Contents

Cardinal-Infante
Ferdinand of Austria at
the Battle of Nördlingen

Rubens (1577-1640)

Achilles discovered
by Ulysses and
Dismedes

Rubens (1577-1640)

Mixed Reality allows to display
spatial-aware content anywhere and anytime



Mixed Reality Contents

Cardinal-Infante
Ferdinand of Austria at
the Battle of Nördlingen

Rubens (1577-1640)

Achilles discovered
by Ulysses and
Dismedes

Rubens (1577-1640)

Problem: How to **align** these **virtual contents** on
precise objects or locations?

Low Overhead Solution?



Book Store



Shopping Center




Museum



Grocery



One thing in common – *Lights!* 

Optical Camera Communication (OCC)

Feature 1 - **Location-awareness**: inherently **links data** to light's **location**

Feature 2 - **Pervasiveness**: LED lights and cameras are pervasive

What if we reuse **Lights** as **transmitters**, and **cameras** as **receivers**...

One thing in common – **Lights!** 



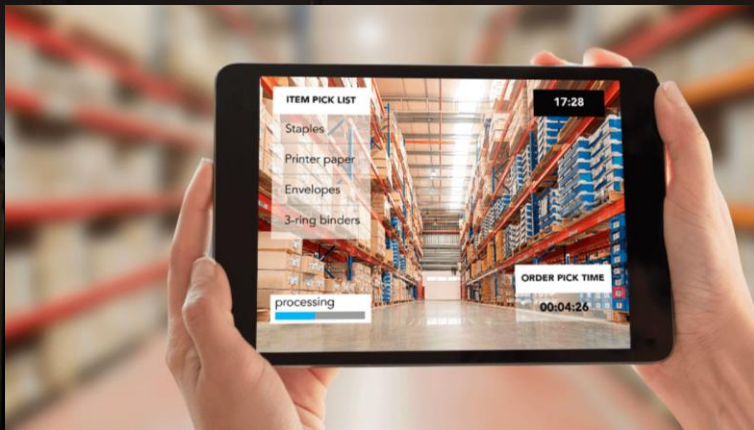
0101010101
01010110...



Optical Camera Communication (OCC)

Feature 1 - *Location-awareness*: inherently *links data* to light's *location*

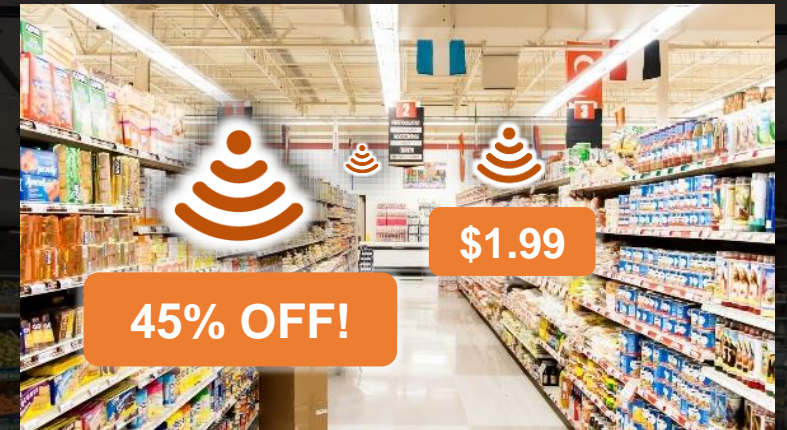
Feature 2 - *Pervasiveness*: LED lights and cameras are pervasive



MR Content
Delivery



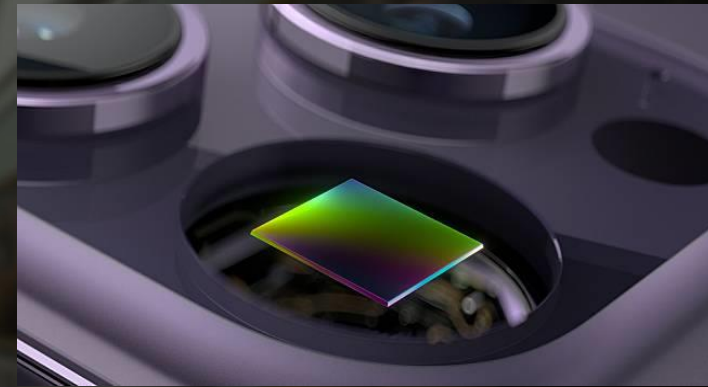
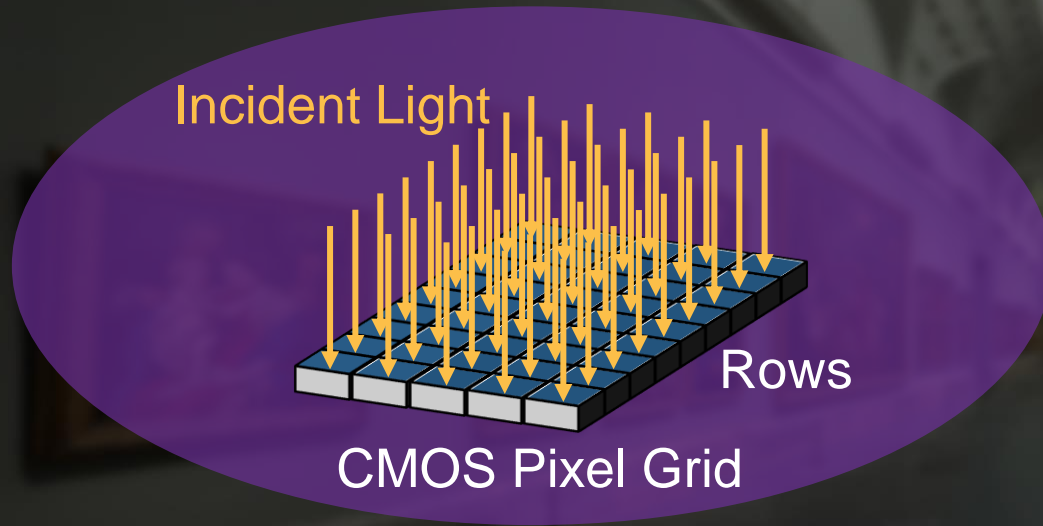
Indoor
Localization



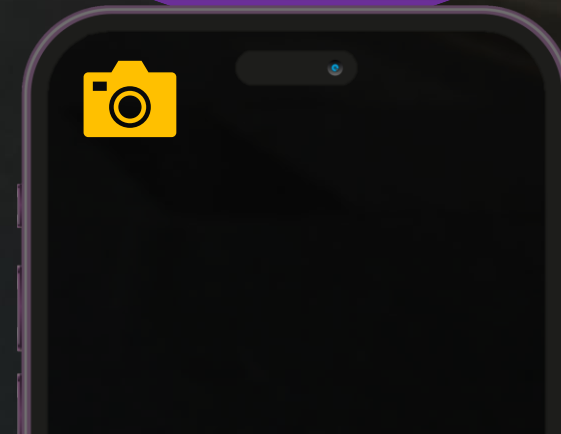
Pervasive
Connectivity

OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*

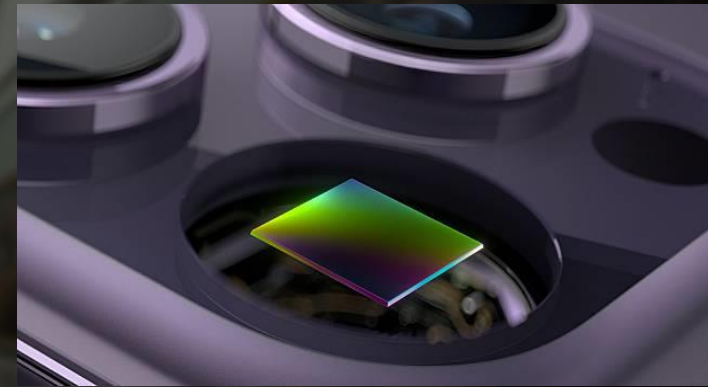
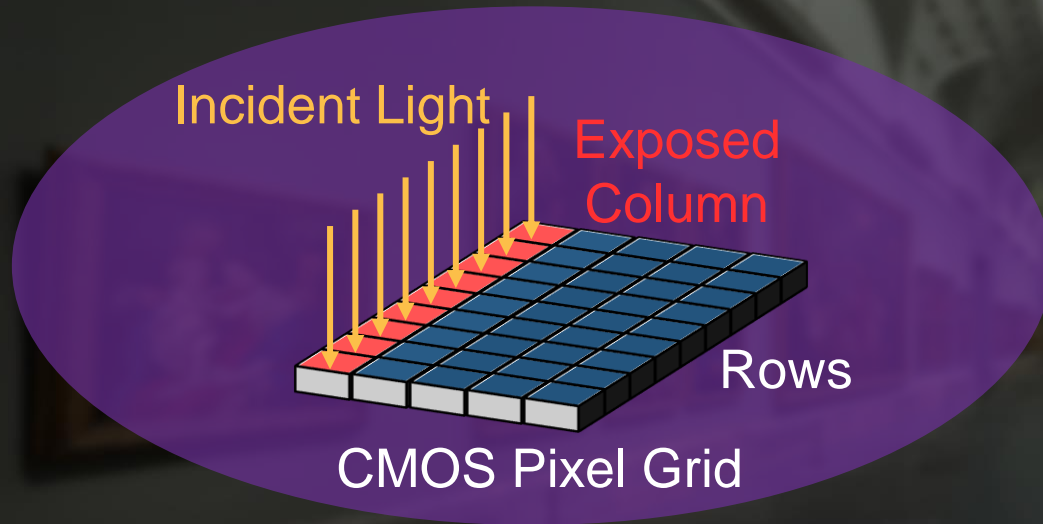


Rolling Shutter CMOS

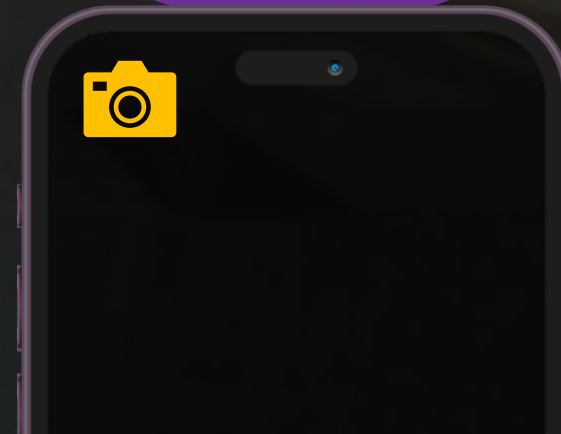


OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*

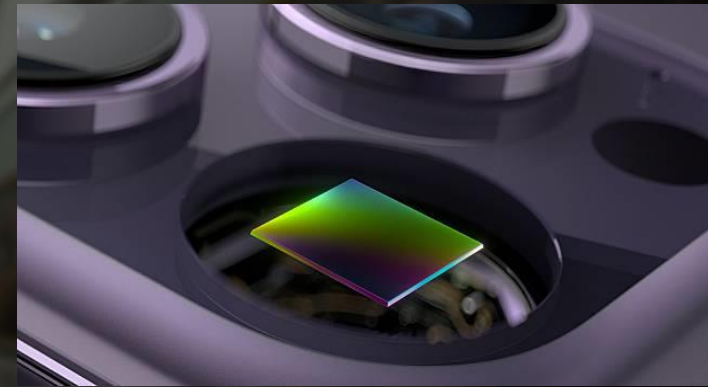
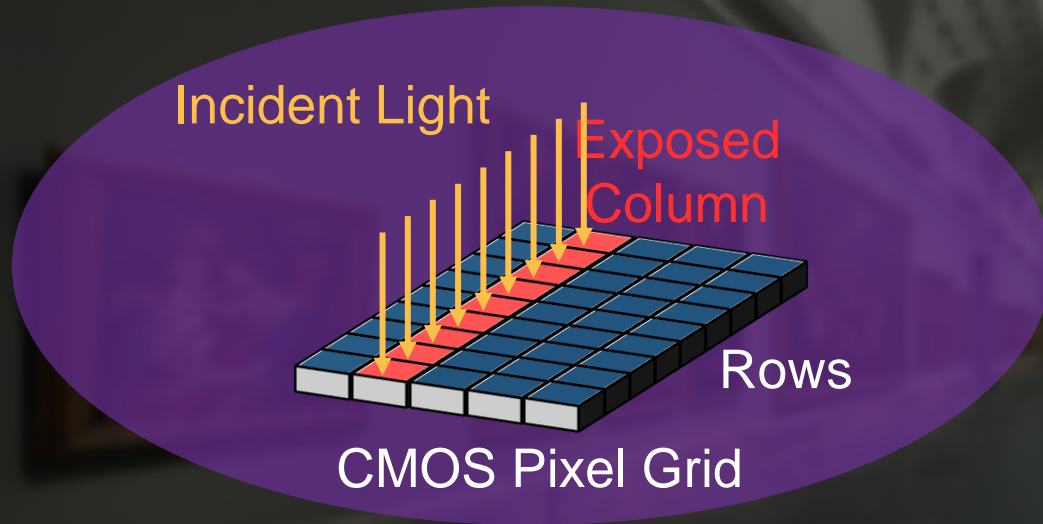


Rolling Shutter CMOS

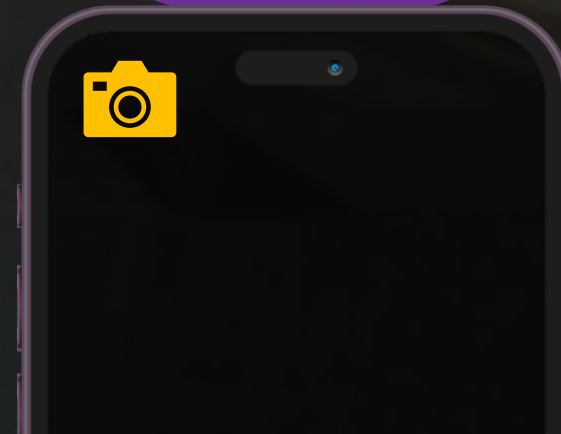


OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*

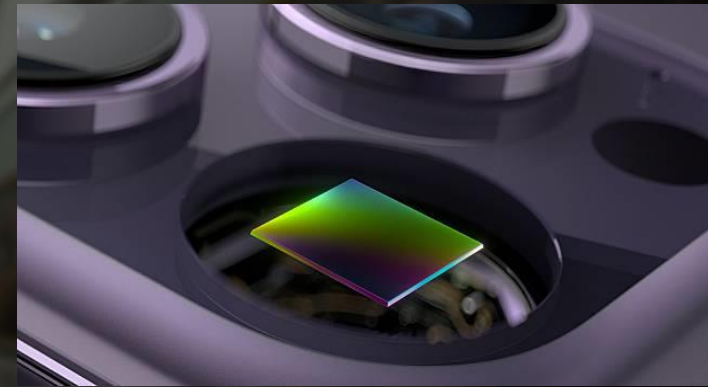
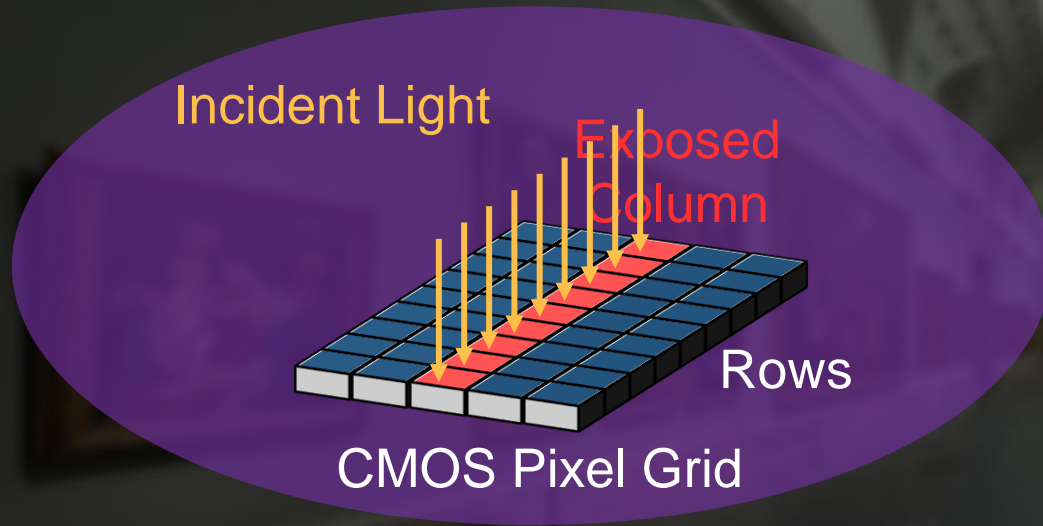


Rolling Shutter CMOS

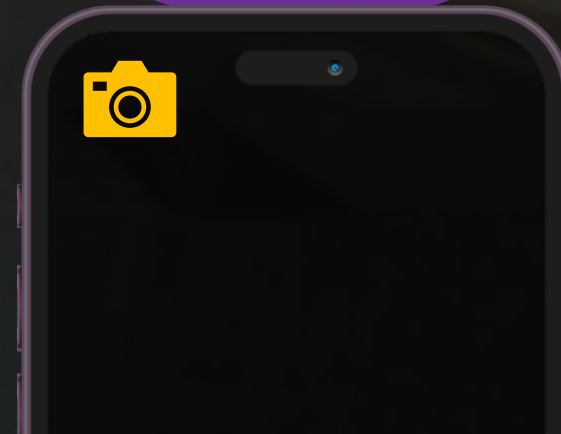


OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*

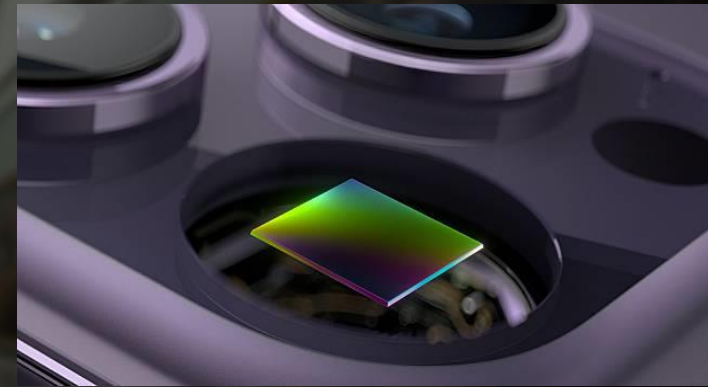
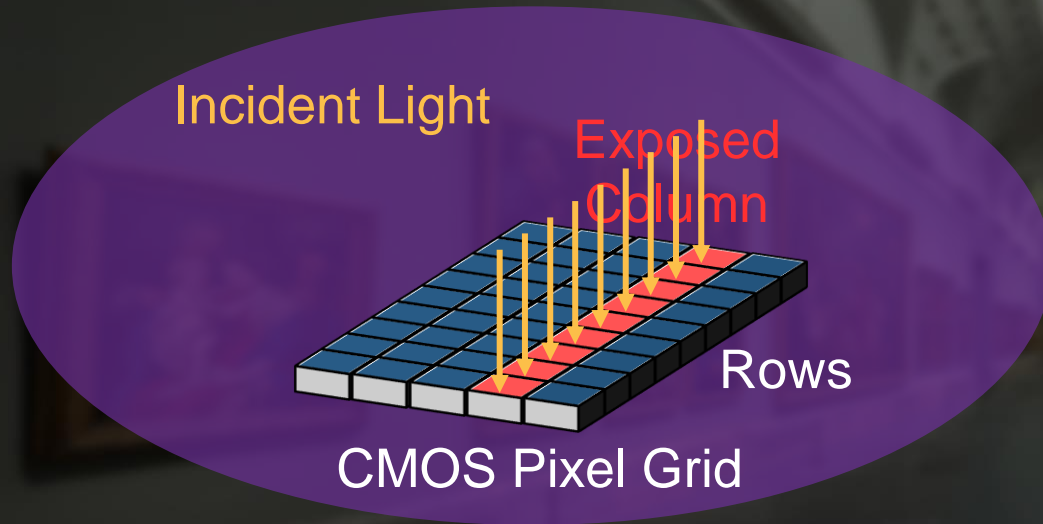


Rolling Shutter CMOS

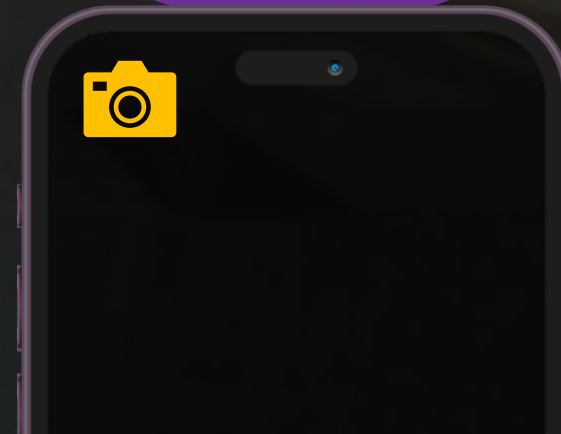


OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*

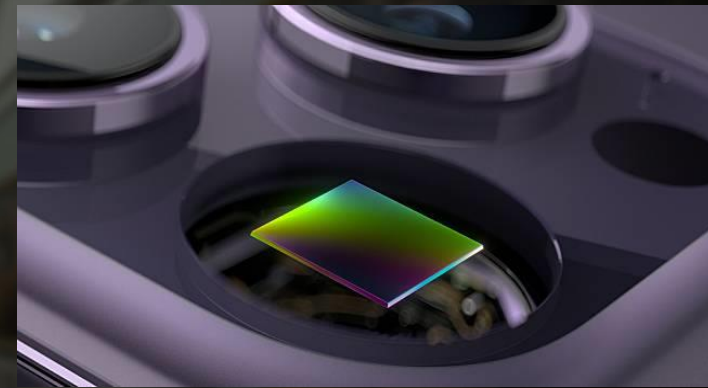
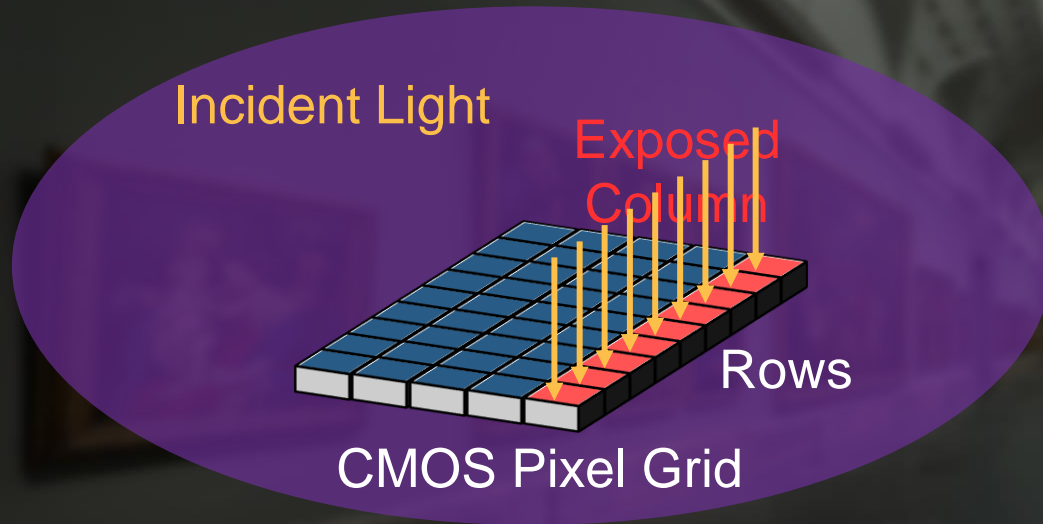


Rolling Shutter CMOS



OCC with Rolling Shutter CMOS

- Rolling shutter exposes a frame *column by column*



Rolling Shutter CMOS



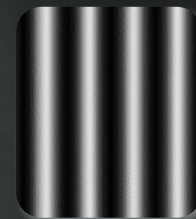
Time

e.g. FSK

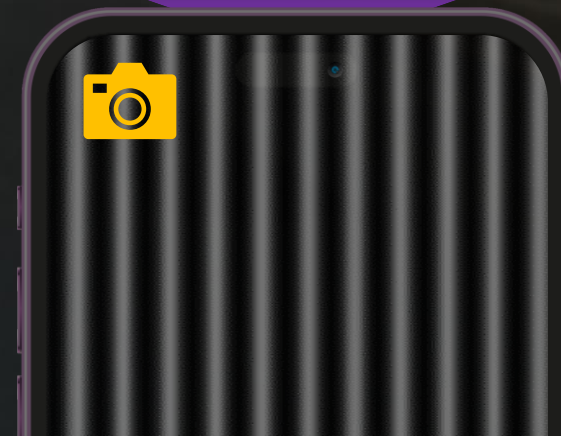
Modulated Stripes



Bit 0



Bit 1



Issues with Existing OCC Designs

Signal-to-Noise Ratio (SNR) $\downarrow \longrightarrow$ Symbol Error Rate (SER) \uparrow

SER < 0.01

SER > 0.01

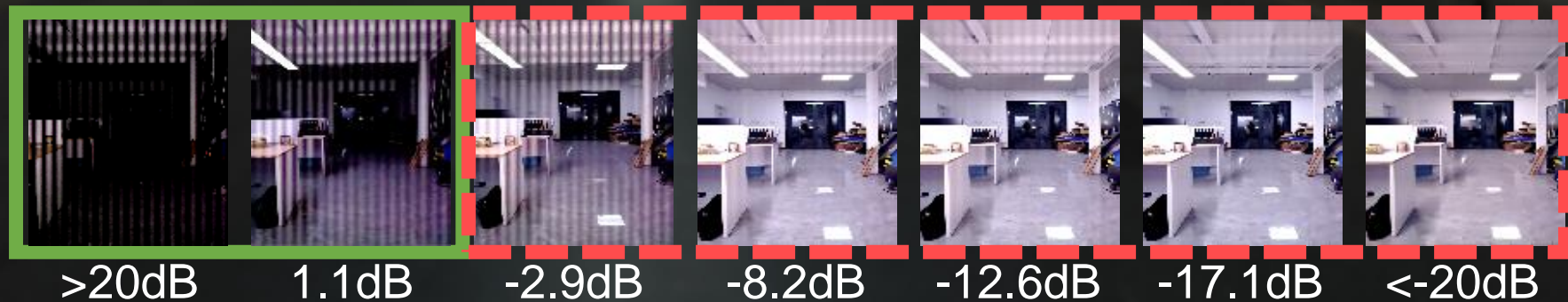
Clean
Background



SER < 0.01

SER > 0.01

Complex
Background



Issues with Existing OCC Designs (Cont.)

1



-12.8dB

low SNR but *clean background*

- Limited distance (~0.4m)
- Often difficult to find such a clean reflector

2



1.1dB

complex background but *high SNR*

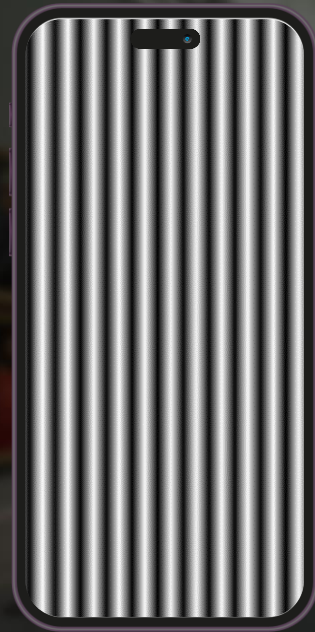
- Still limited distance (~1.4m)
- Pronounced video degradation

WinkLink!

Can we build a novel OCC system that works under *low SNR* with any *unseen complex backgrounds* ?

Issues with Existing OCC Designs (Cont.)

SNR: 20 dB



Traditional OCC

vs.

SNR: -20 dB



WinkLink

Under
Low SNR



Complex
Background



WinkLink!

Can we build a novel OCC system that works under *low SNR* with any *unseen complex backgrounds* ?

Challenge 1: Dynamic Background Entanglement

- The entanglement can be modeled as:



Video Frame I

(what we capture)

=



Background B

+



OCC Signals O

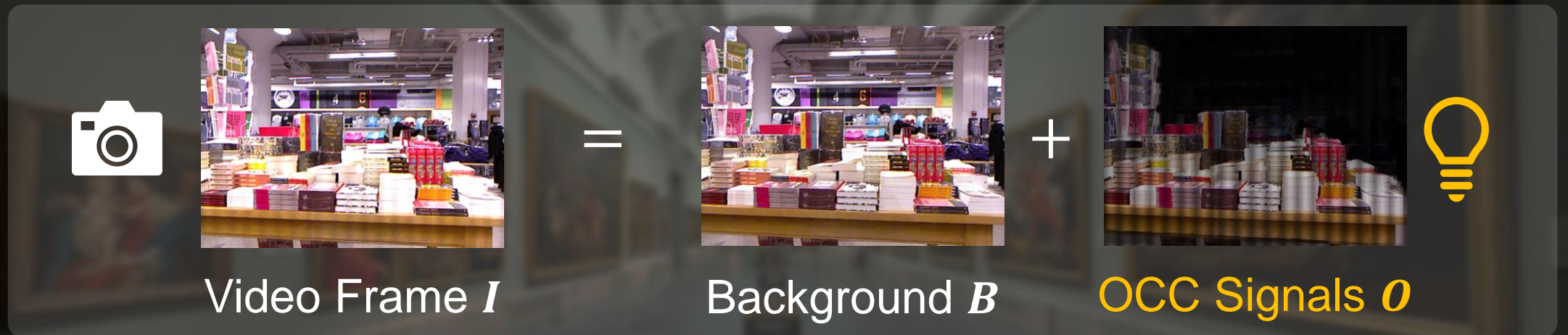
(what we want)



- Disentangling Signal O is an *ill-posed* problem.
- The background B is *dynamic*, varying across different frame I .

Solution: DNN-based Signal Extraction

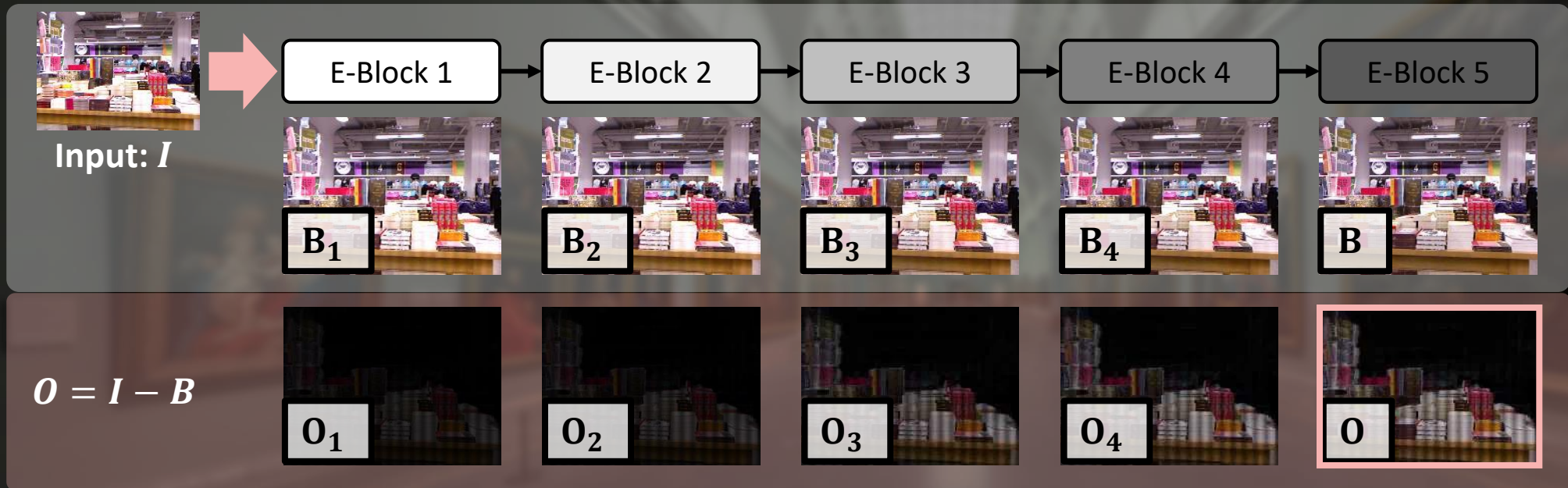
- The entanglement can be modeled as:



- Key insight:
 - DNN can handle ill-posed problem*** by implicitly enforcing constraints
 - Replication of signals across rows \rightarrow Spatial correlation
(DNNs excel at capturing spatial correlation)

Solution: DNN-based Signal Extraction

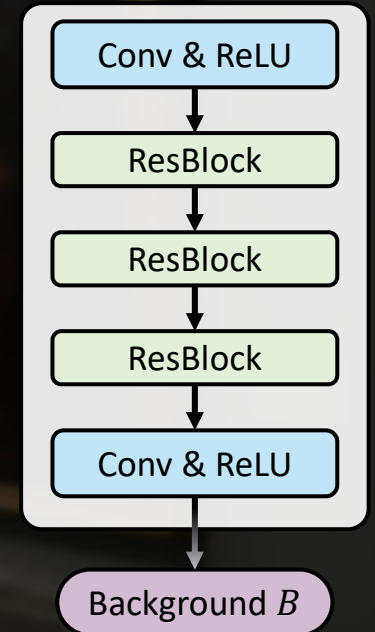
Stage I: Progressive Signal Extraction



not fully disentangled

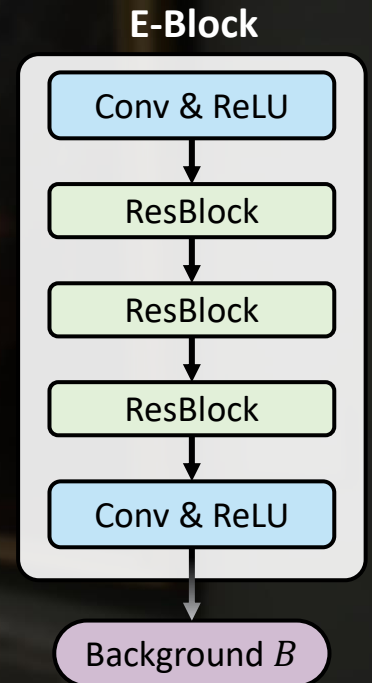
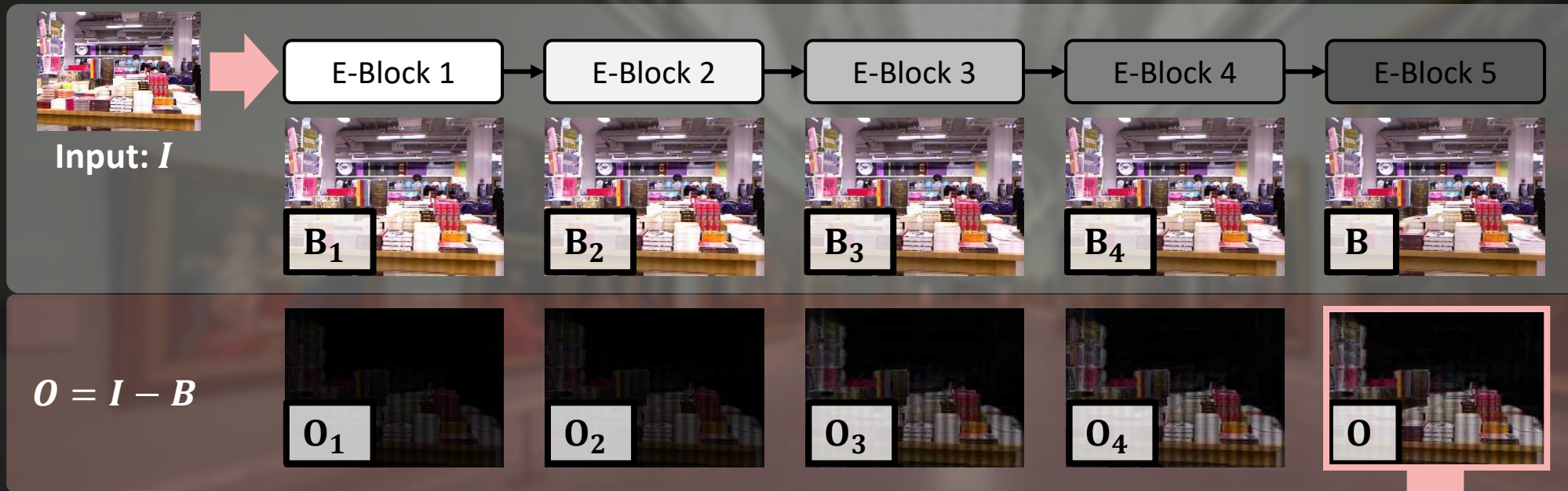
well disentangled

E-Block

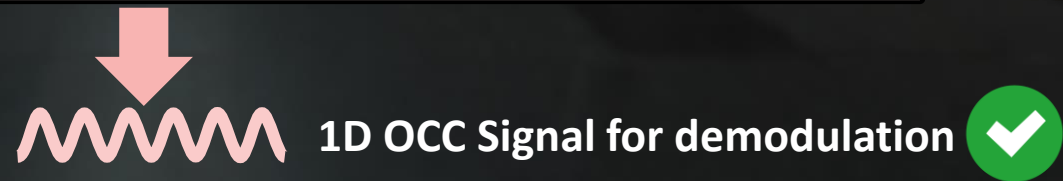


Solution: DNN-based Signal Extraction

Stage I: Progressive Signal Extraction



Stage II: Signal Fusion Across Rows



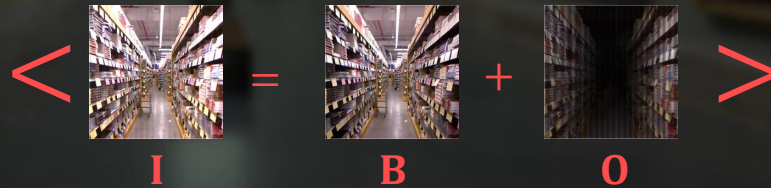
Challenge 2: Laborious Training Data Preparation

Generalizability of WinkLink



Diversity of Training Data

- 1. Unseen backgrounds
- 2. Diverse stripe patterns
- 3. Varying SNRs



- We require a **large dataset** of **paired** $\langle I, B, O \rangle$ with above diversities.
- Manual assembly of such a diverse dataset is **time-intensive** and **impractical**.

Solution: Scalable Training Data Synthesis

$$\begin{matrix} < & \text{I} & = & \text{B} & + & \text{O} & > \end{matrix}$$
The diagram illustrates the synthesis of a target image I. It shows three square images arranged horizontally, separated by mathematical symbols. The first image, labeled 'I' below it, is a bright, clear view of a library aisle with tall bookshelves on both sides. The second image, labeled 'B' below it, is a similar view of the same library aisle, also bright and clear. The third image, labeled 'O' below it, is a dark, almost black image of the same library aisle, representing a light attenuation or shadow map. The images are connected by the symbols '<', '=', '+', and '>' in a sequence that reads '< I = B + O >'.

I **B** **O**

- Target: synthesize paired $< I, B, O >$ -- *easy to scale*
while *minimizing the gap* between synthetic and real data
- **Key components** for precise synthesis:
 - the *light reflection* model under the Lambertian assumption
 - the *light attenuation* on varying distances

Solution: Scalable Training Data Synthesis

- Input:



Depth map
(only for training)



Background B

- Output:



Frame I



Signal 0

Solution: Scalable Training Data Synthesis

- Input:



Depth map
(only for training)



Background B



Background B

- Output:



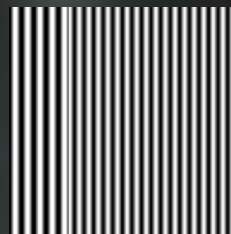
Frame I



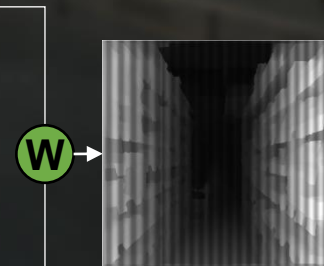
Signal O



Depth map



Random
Stripe



Illumination
from OCC

-- modeling light attenuation

\textcircled{W} : Weight by Depth

Solution: Scalable Training Data Synthesis

- Input:



Depth map
(only for training)



Background B

- Output:



Frame I



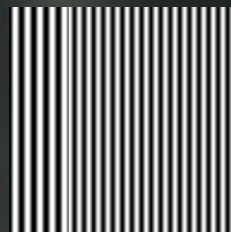
Signal O



Background B



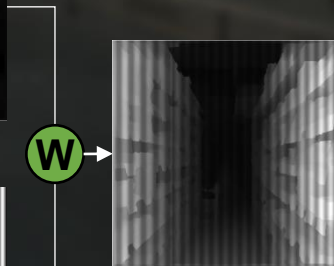
Depth map



Random
Stripe



superimpose
under **Lambertian assumption**
 $I = \text{Reflectance} \times \text{Illumination}$



W



Illumination
from OCC

W : Weight by Depth

Solution: Scalable Training Data Synthesis

- Input:



Depth map
(only for training)



Background B

- Output:

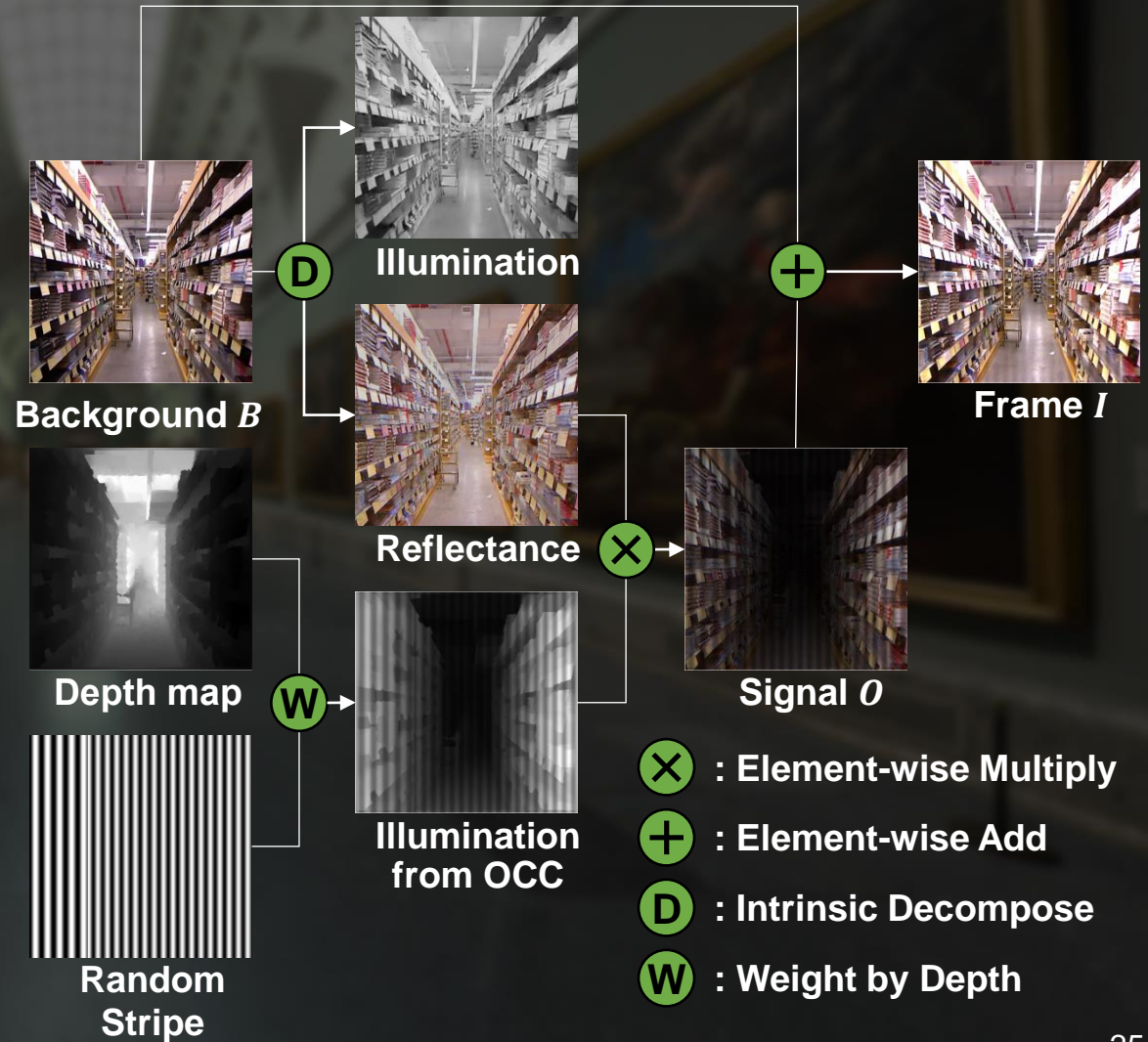


Frame I

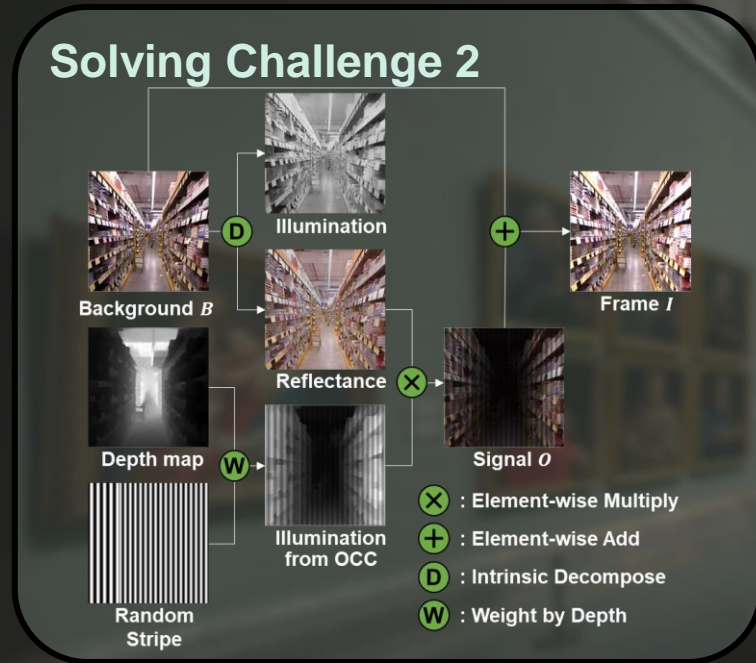


Signal O

- Source dataset: NYU Depth V2 dataset
- We synthesize **7245** frames. ✓



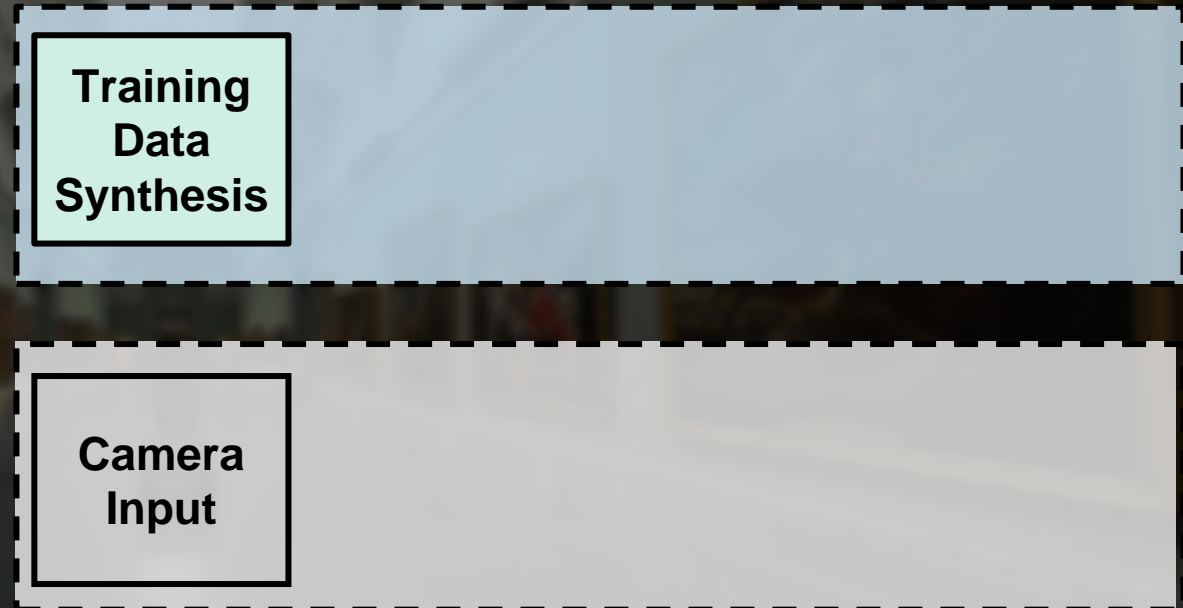
Design Overview of *WinkLink*



Transmitter



■ Bootstrapping ■ Deployment

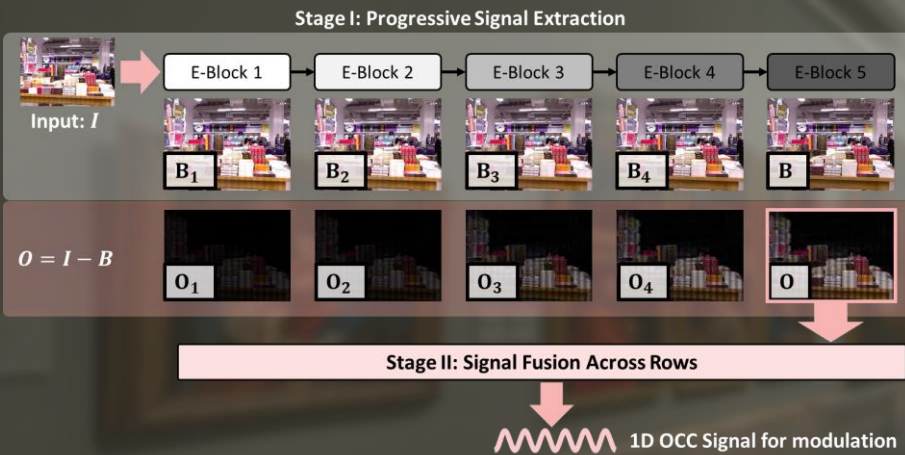


Receiver



Design Overview of *WinkLink*

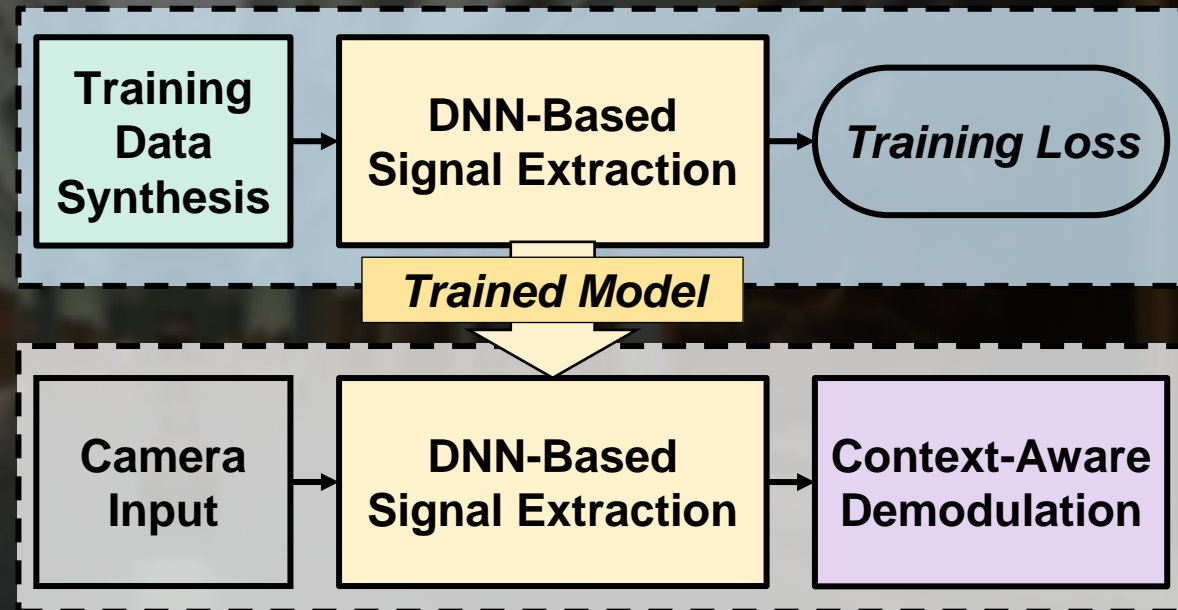
Solving Challenge 1



Transmitter



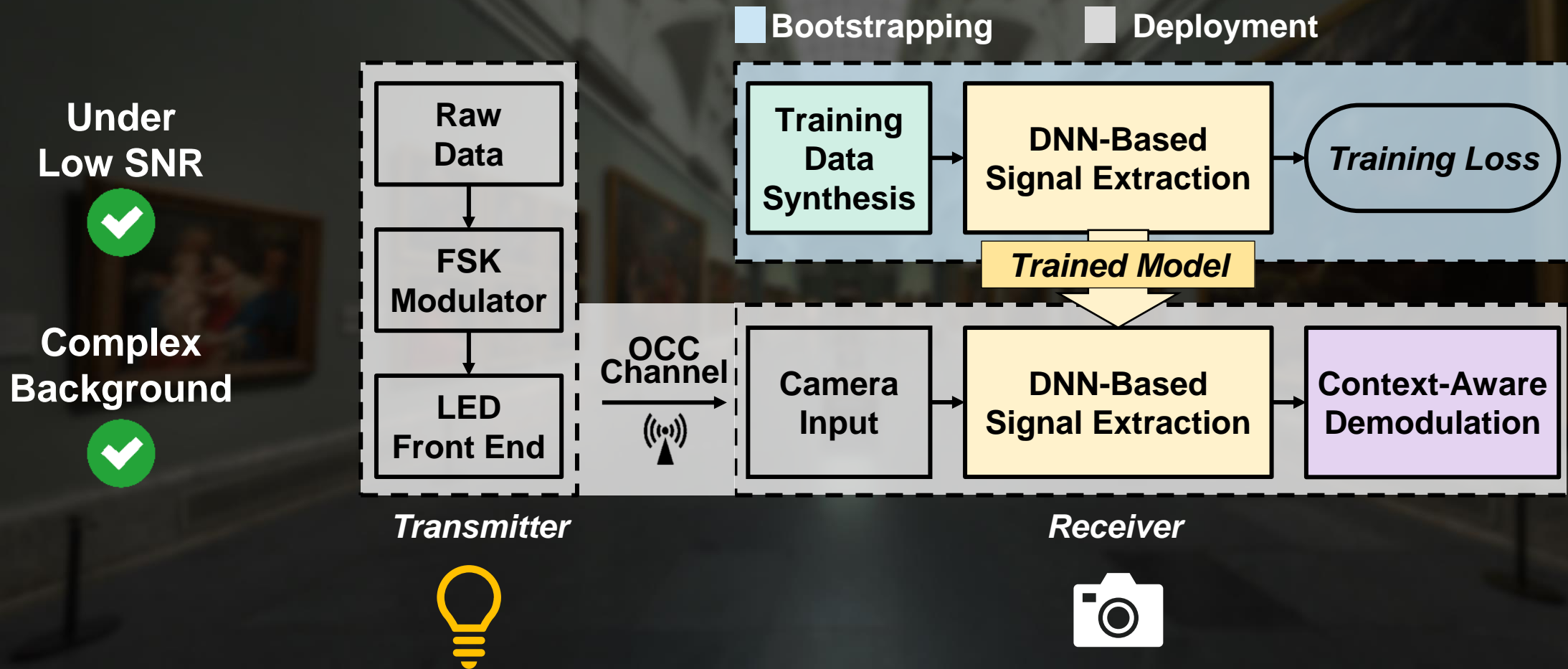
■ Bootstrapping ■ Deployment



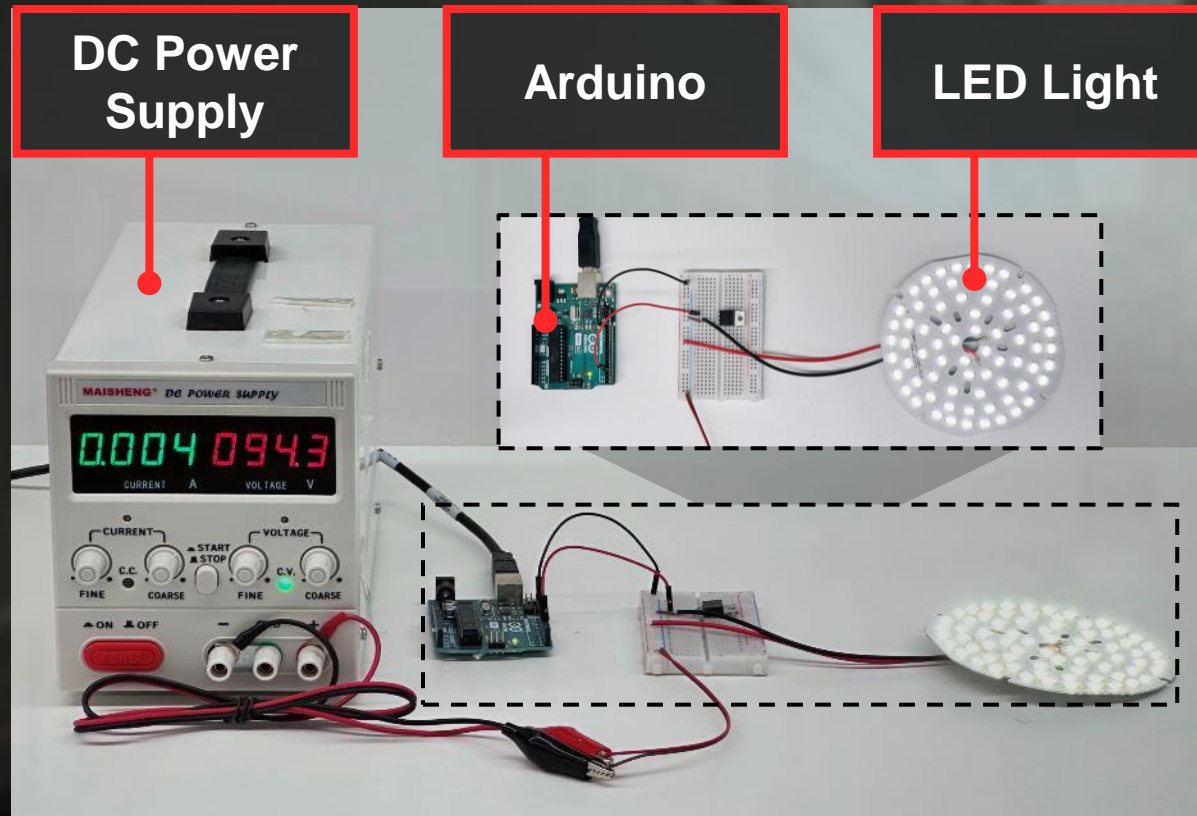
Receiver



Design Overview of *WinkLink*



Prototype and Implementation

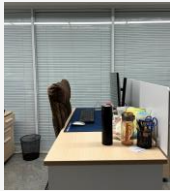


- Transmitter
 - 5 watt LED
 - Modulation: 4-FSK
 - Each symbol: 1/60 seconds (120 bps)
- Receiver
 - Phone (iPhone/Samsung/Huawei)
 - Frame Rate: 60 FPS
 - Resolution: 512x512

Overall Performance



Office



Glass



Stair



Ornament



Canteen



Cafe



Game Station



Posters



Bookshelf



Plants



Workshop

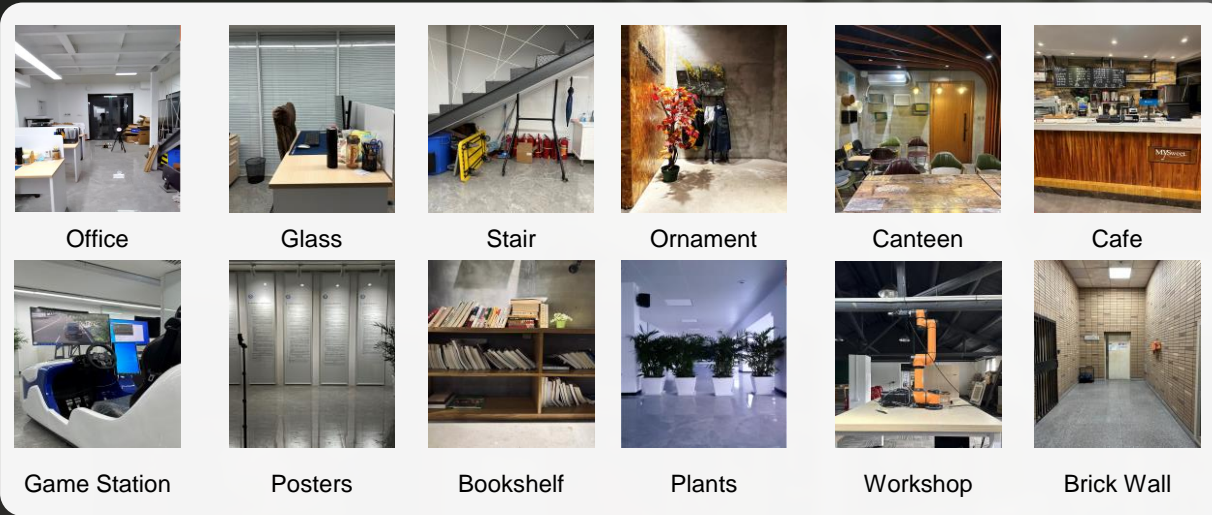


Brick Wall

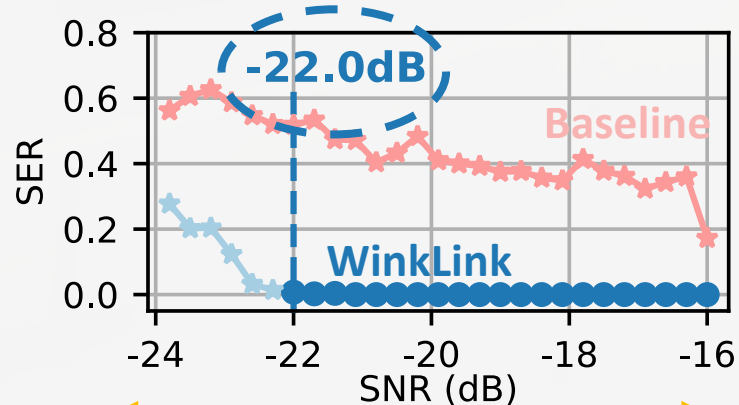
- Test Dataset

- 12 unseen environments
- 30K frames per environment
- ***genuinely captured and not synthetic***

Overall Performance



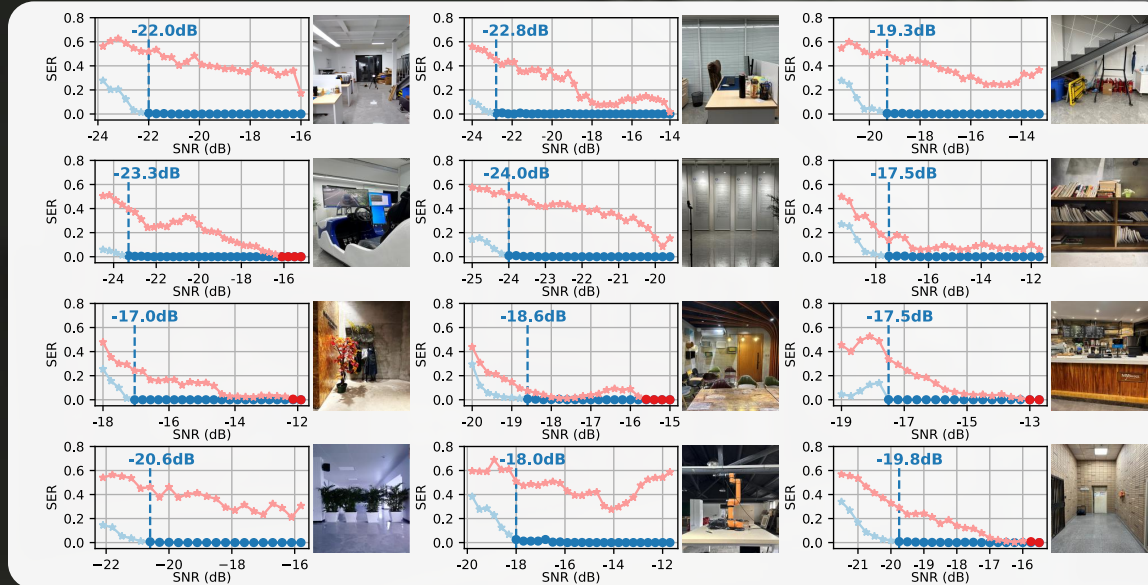
- Test Dataset
 - 12 unseen environments
 - 30K frames per environment
 - *genuinely captured* and *not synthetic*
 - **SNR variation** by adjusting **LED power** from 0 to 5 watts
- WinkLink vs. Baseline
- Metric: **mSNR** – minimum SNR at which SER drops below 0.01



Office

Environment 1/12

Overall Performance



➤ Results

- **WinkLink** shows an average mSNR of **-20 dB**
- Consistently outperforms baseline with a **5.8 dB** SNR gain

• Test Dataset

- 12 unseen environments
- 30K frames per environment
 - *genuinely captured* and *not synthetic*
 - **SNR variation** by adjusting **LED power** from 0 to 5 watts

• WinkLink vs. Baseline

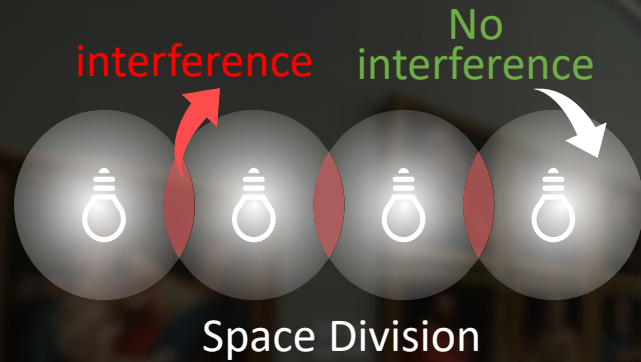
- Metric: **mSNR** – minimum SNR at which SER drops below 0.01

Summary of Other Evaluation Results

- Works with **three distinct phone models**: iPhone 14 Pro, Huawei P40 Pro, and Samsung Galaxy S21
- Performs well when the user is **moving** at speed of 2m/s (under **dynamic backgrounds**)
- Works at a distance up to **11 meters** with a **10 watt LED**
- Has **minimized interference** on concurrent vision applications (e.g., object detection)

Discussion

- Interference between Multiple OCC Links



Light steering



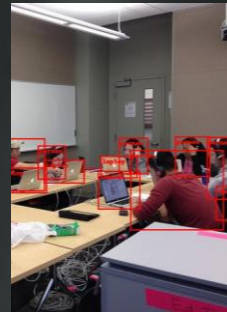
- Integrated Sensing and Communication on Vision



Sensing with Vision



Segmentation



Detection



**Communication
with WinkLink**

100101011101010010 ...

Conclusion

- We propose WinkLink, a novel OCC system that operates under **unseen and complex backgrounds**, while maintaining **low-SNR requirements**.
- We hope to explore the **integrated sensing and communication** on vision domain.



Thank you!



Homepage

Mail: ruixiao24@zju.edu.cn



Rui Xiao

PhD candidate at Zhejiang University

Mobile, Wireless, Sensing, Security

I'm seeking a **post-doctoral position** starting in **Fall 2025**.
Please feel free to contact me!