#### DeepLight: Robust & Unobtrusive Real-time Screen-Camera Communication for Real-World Displays

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#### Screen-Camera Communication: Background

Msg. to be encoded: "So she was considering ..."

Decoded Msg.: "So she was considering ...."

> Demodulate camera frames

Modulate video frames

**Screen Camera Visual Channel** 

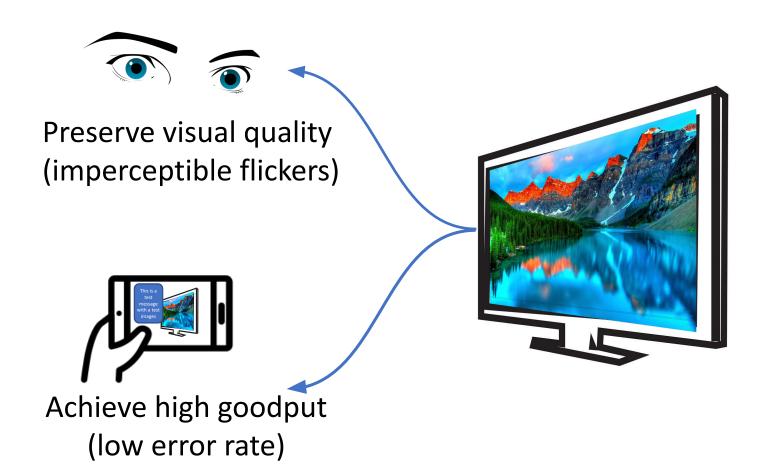




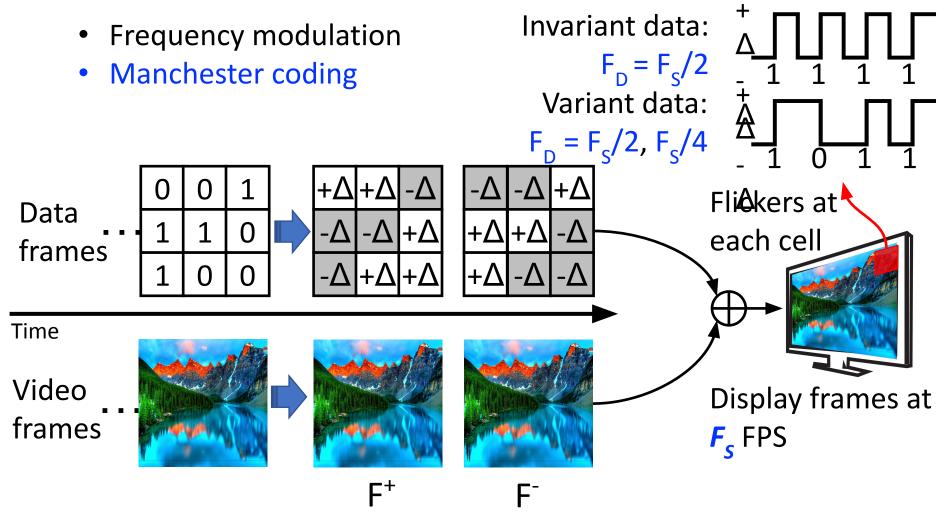
Augmented text, instructions, audio, ... on public screens

- Only users, who want to, receive hidden information.
- Avoid annoying other users

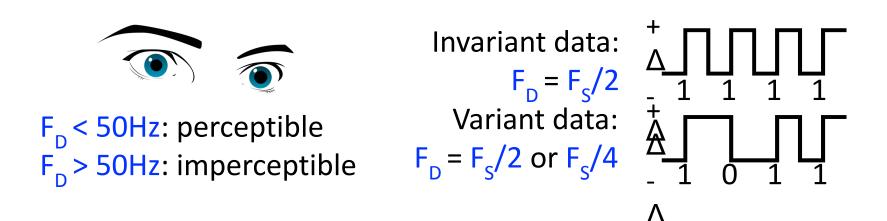
#### Two key objectives



#### State-of-the-art Encoder: How hidden data is embedded?



### State-of-the-art Encoder: Suppress flickers with high display rates

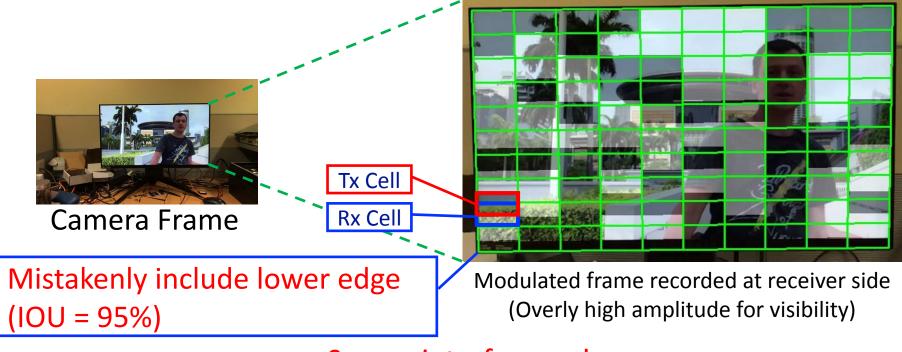


To achieve imperceptible flickers:

- F<sub>s</sub> > 100 FPS for invariant data
- F<sub>s</sub> > 200 FPS for variant data

How to support imperceptibility at common frame rates (e.g., 30, 60FPS) ?

# State-of-the-art Decoder: Grid splitting



Severe interference!

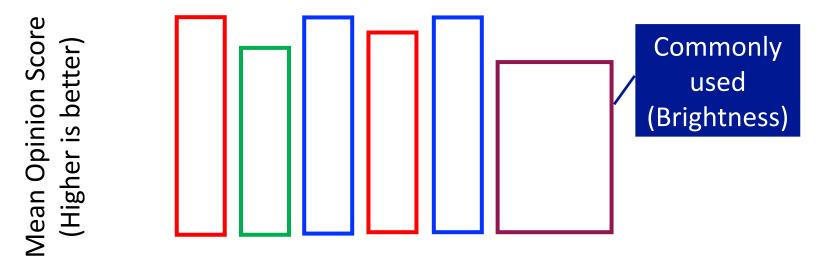
How to support robust decoding with imperfect screen extraction?

#### DeepLight contributions

- Blue channel modulation for imperceptibility at common frame rates (60 FPS)
- A holistic decoding method using convolutional neural network; support imperfect screen extraction
- A hybrid screen extraction method for practically high screen extraction accuracy

#### Imperceptibility at **common** (60FPS) display rates: Blue light

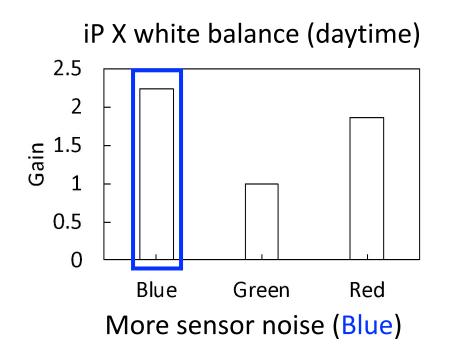
Human eyes are known to be less sensitive to Blue color



Modulation channel & amplitude ( $\pm\Delta$ )

- Green, Brightness: low MOS even with the lowest amplitude (±1)
- Red: Low visual quality with higher amplitude (±2)
- **Blue**: High visual quality even with higher amplitude (±2)

#### Cope with noise





Cross channel noise (Imperfect screen extraction)

Increase mod. amplitude  $\Delta$ ? X Decrease visual quality

- Learning-based decoding instead of hard thresholding
- Avoid grid-splitting

#### DeepLight Holistic Decoder

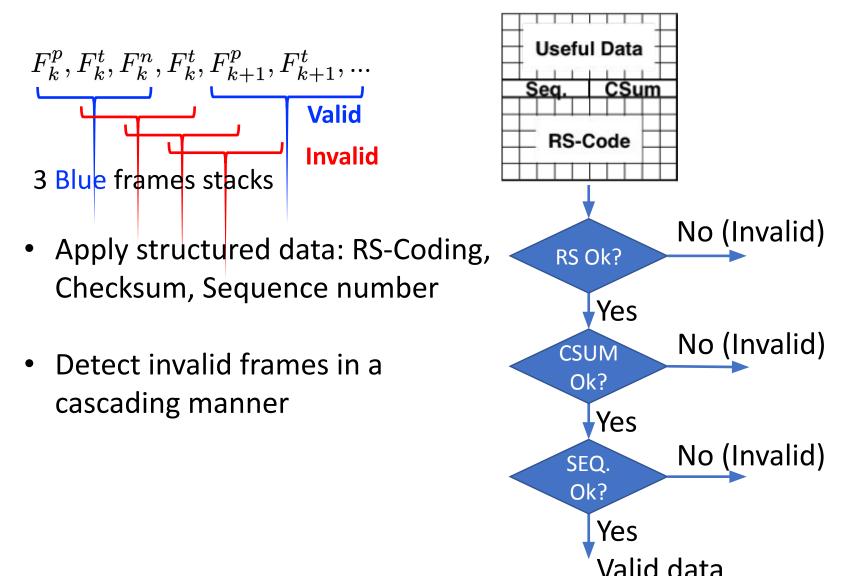
299 5x5x100 1x1x16 3x3x1 Input Image 299 SNN CNN CNN 9 9 x x 32 Learn function output = 13 Blue frames  $(F_k^{\ n} E a_k^{\ t} F_k^{\ n})$  is inferred using the entire "screen", not just a cell  $F_{k}^{p}, F_{k}^{t}, F_{k}^{n}, F_{k}^{t}, F_{k+1}^{p}, F_{k+1}^{t}, \dots$ 

Learn temporal relation (Manchester coding)

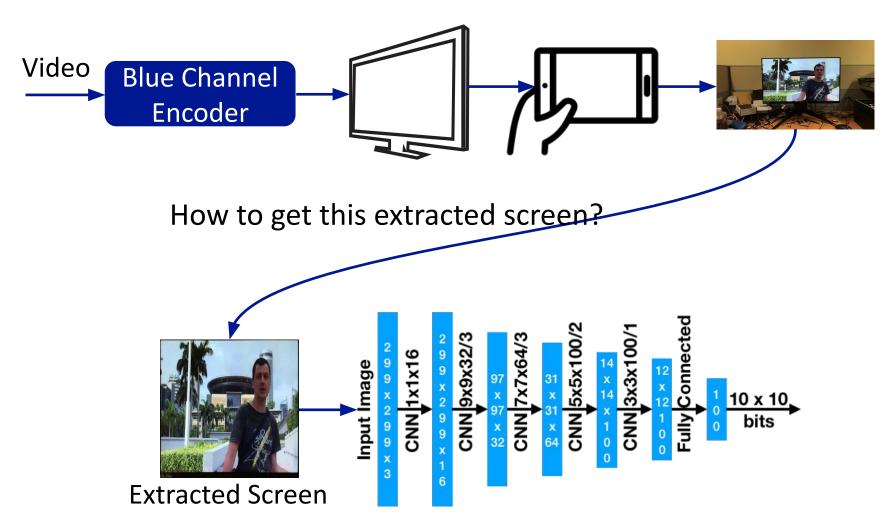
Assume  $F_{camera} = 2F_{display}$  $F_{\mu}^{s}$ : Camera frame corresponding to Manchester pair k frame type s (positive, transition, negative)

bits

#### Filtering out invalid frames

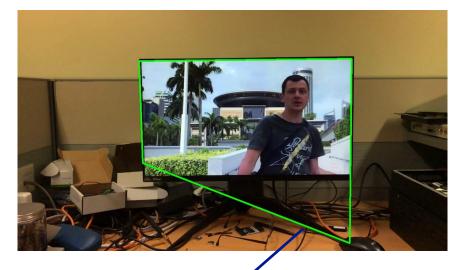


#### DeepLight screen detection



#### DeepLight screen detection

"practically" accurate screen extraction is still necessary



Canny + Hough Transform: Tricked by nearby "line" textures

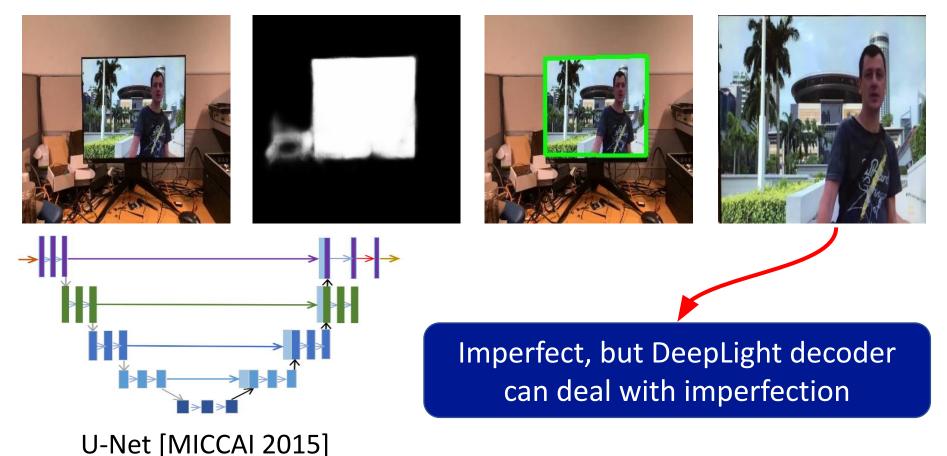


#### Expected



#### DeepLight screen detection

1. U-net based segmentation: Filter out "non-screen" areas 2. Contour analysis 3. Perspective Transform



### DeepLight Evaluation

Mean Opinion Score (MOS):

- 1: Very unpleasant
- 2: It's bad
- 3: It could be better
- 4: It's good
- 5: Cannot differentiate from the original video

Performance metrics:

- Raw throughput: (1-BER)\*F<sub>s</sub>\*D
- Throughput: (1-FER) \*F<sub>s</sub>\*D
- Goodput: (1-FER) \*F<sub>s</sub>\*U

Note: Raw throughput is not informative if BER is high

Default settings:

- We used a 25" monitor
- Display rate: 60 FPS
- Grid size: 10x10

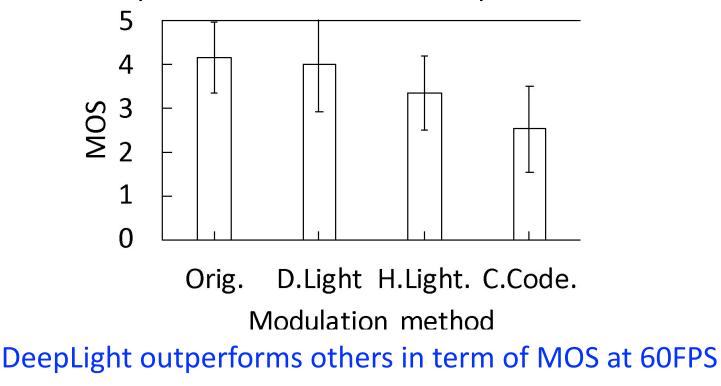
**BER**: Bit error Rate; **FER**: Frame error rate (only recoverable frames)

**D**: Number of bits in a frame; **U**: Number of useful bits in a frame; **F**<sub>s</sub>: Display frame rate

#### DeepLight preserves visual quality

17 participants:

- 10 males, 18 to 32 years old
- 1 astigmatism, 3 farsighted, 7 shortsighted
- Each person watches 6 video clips X 4 versions



D.Light: DeepLight, H.Light: HiLight, C.Code: ChromaCode

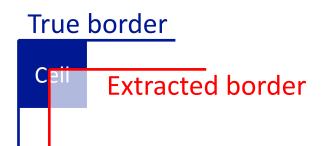
## DeepLight performance with fixed camera

0% screen extraction error

> 0% screen extraction error (d=1.5m)

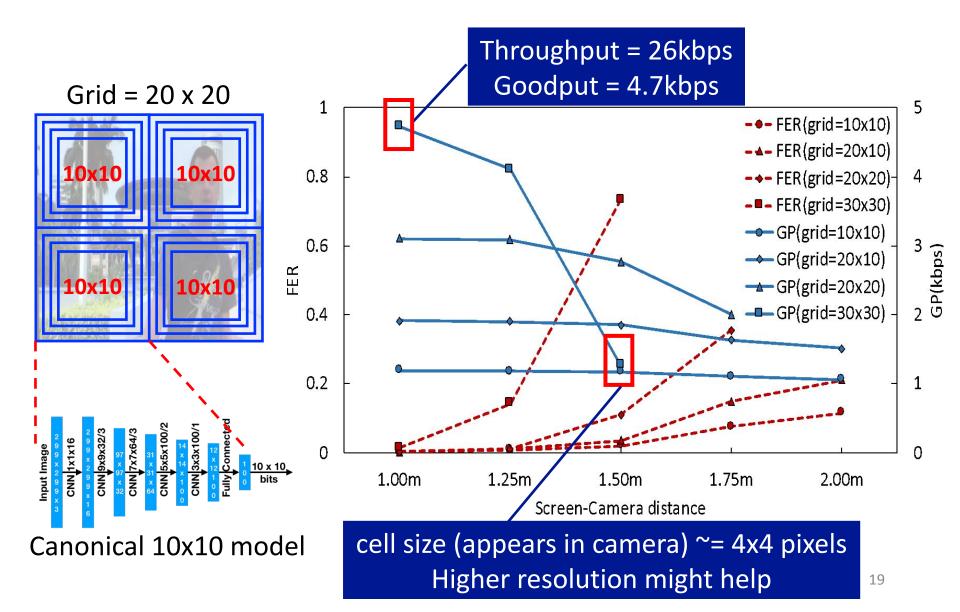
> 1.0 Kbps even at 2.0m

Example 40% SHF error: 64% cell area loss

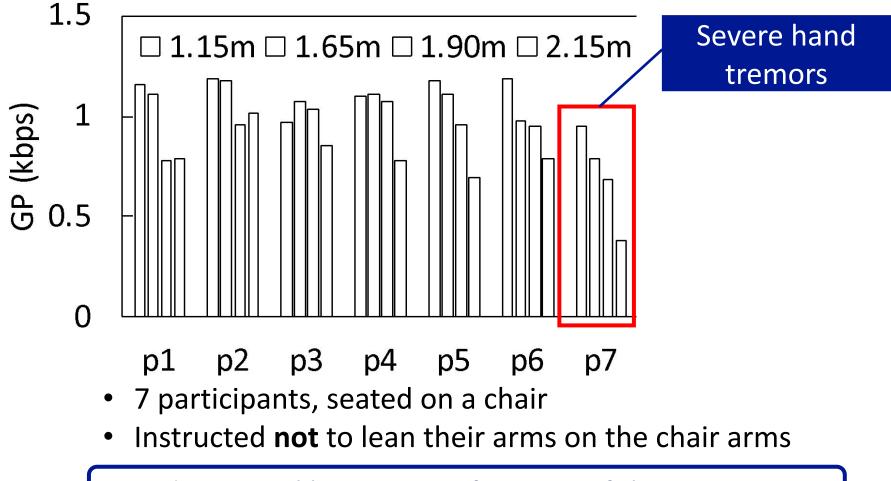


SHF: Shift, EXP: Expand, SHR: Shrink, ROT: Rotate

#### Support larger grid size

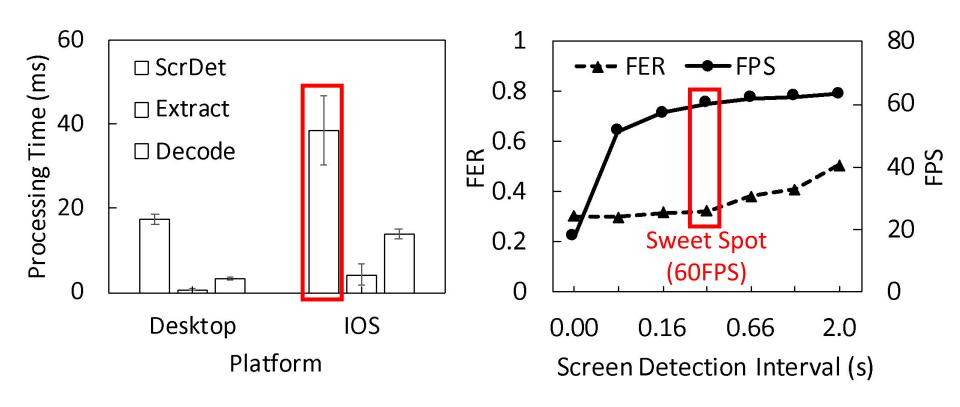


## DeepLight performance with hand-held camera



Goodput > 0.9kbps at 1.9m for most of the participants

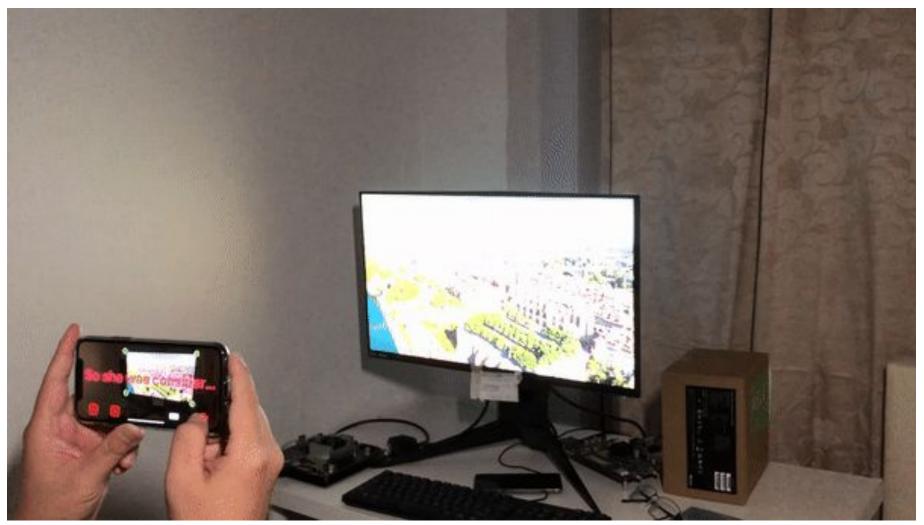
#### DeepLight on Smartphone



 We do not need screen detection for every camera frame (8.3ms)

Experiment with a **walking** user (more motion artifacts)

#### DeepLight captioning app.



- Press to process the latest 32 frames in buffer
- Detect screen in the first frame only

#### Comparison with previous works

*	·			·	
Work	Require SCR locator	Visual Quality	Throughput (kbps)	Goodput (kbps)	Processing time on mobile phone (ms)
DeepLight	No	Very high	26.6	4.7	16.6 (iPhone 11 Pro)
ChromaCode (2018) [5]	Yes (Black & White lines)	Low	N.A. (220 Raw throughput)	0.5 – 1.5	500 (Pixel2)
TextureCode (2016) [8]	N.R (Offline extracted)	N.A.	11.25	N.A.	N.A. (Offline)
Inframe++ (2015) [3]	Yes (QRCode locator)	N.A.	9	N.A.	200 (Core-i5 CPU + FirePro V3900 GPU)
Hilight (2015) [4]	Yes (OFF/ON screen)	High	4.6	N.A.	5 (iPhone 5)

Some values are borrowed from [5] and [8], normalized to 60FPS

DeepLight code: <u>https://github.com/LARC-CMU-SMU/deeplight</u>



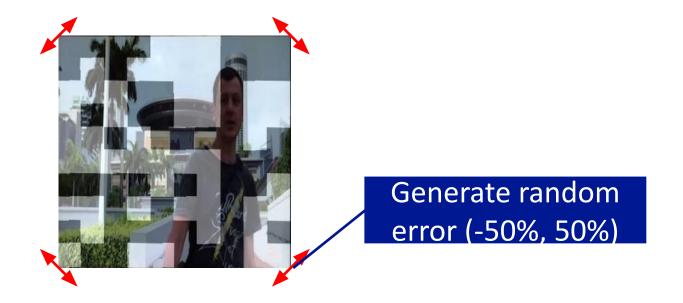
#### Summary

- Apply Blue channel modulation for imperceptibility at common frame rate (60FPS)
- Develop a hybrid (U-Net + classic contour analysis) screen extraction method for practically accuracy
- Develop a CNN-based holistic decoder that support robust decoding with imperfect screen extraction
- Collectively, DeepLight is robust enough to support hand-held camera and mobile execution

Thank you!

#### More details ...

#### Training LightNet



- Stack 3 consecutive Blue frames to form a sample
- Sample  $S_k = \{B_k^{p}, B_k^{t}, B_k^{n}\}$
- 22500 fixed camera samples
- 25200 hand-held camera samples

#### Training ScreenNet

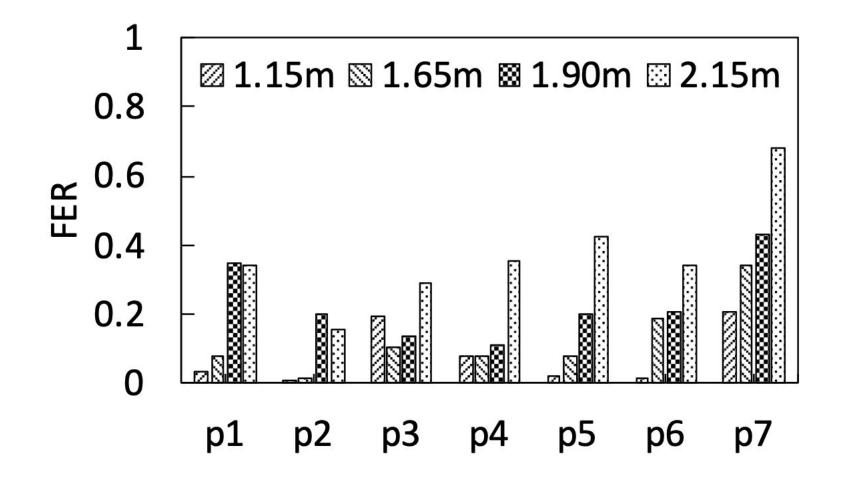


- Collect screens from Google search
- Take photos of screens at different places
- ~800 images + data augmentation (rotation, displacement, scale, ...)

#### ScreenNet performance

	Kernel size		
	$1 \times 1$	$2 \times 2$	$3 \times 3$
Indoor	0.93/0.95	0.89/0.97	0.83/0.99
Outdoor	0.83/0.97	0.82/0.97	0.80/0.94

#### Hand-held performance



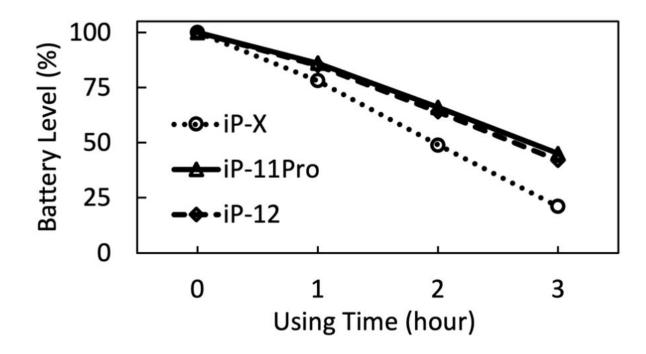
#### DeepLight vs. Viewing angle

	Distance [m]		
Viewing Angle	1.00	1.50	2.00
0°	0.4/ <b>1.2</b>	2.1/ <b>1.18</b>	11.7/ <b>1.06</b>
15°	0.3/1.2	1.2/ <b>1.19</b>	9.3/ <b>1.09</b>
30°	0.2/1.2	1.9/ <b>1.18</b>	6.3/ <b>1.12</b>
45°	5.6/ <b>1.13</b>	14.8/ <b>1.02</b>	30.6/ <b>0.83</b>
60°	76.1/ <b>0.29</b>	94.5/ <b>0.07</b>	100.0/ <b>0.0</b>

#### DeepLight vs. Ambient lighting

Lighting	FER/GP
eFL+BG	2.8/1.17
eFL+LED	1.4/ <b>1.18</b>
iFL+LED	6.5/ <b>1.12</b>
iFL	9.7/ <b>1.08</b>

#### **Energy consumption**



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