

Electronic Eyes Using Flexible and Neuromorphic Optoelectronics

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Introduction



Bio-inspired Optoelectronics based on flexible/synaptic devices and nanomaterials (2D materials, silicon nanomembrane) will enables Machine Vision Applications. **Bio-inspired camera**

In-sensor processing



ACS Nano 18, 1241 (2024) InfoMat 6, e12479 (2024) Adv. Sci. 10, 2304039 (2023) Sci. Adv. 8, eabq3101 (2022) Nature Commun. 11, 5934 (2020)

Flexible/Synaptic device



ACS Omega 8, 5209 (2023) ACS Nano 17, 20013 (2023) Nano Lett. 21, 9153 (2021) Nano Lett. 19, 2741 (2019) Adv. Healthc. Mater. 5, 80 (2016) ChemNanoMat 2, 1006 (2016) Adv. Funct. Mater. 25, 7109 (2015) Nature Commun. 5, 5747 (2014)

Bio-inspired Optoelectronics





IEEE JSTQE 30, 3800911 (2024) IEEE J-FLEX, 1, 76 (2022) Nature Electron. 3, 546 (2020) Adv. Funct. Mater. 28, 1705202 (2018) Nature Commun. 8, 1664 (2017)

Functional Nanomaterials



Nano Lett. 23, 3344 (2023) Adv. Funct. Mater. 33, 2210367 (2023) Acc. Chem. Res. 52, 73 (2019) Adv. Mater. 29, 1702902 (2017) Adv. Mater. 28, 9326 (2016) Nature Commun. 6, 7149 (2015)

From C. Choi's Publications Blue indicates corresponding-author/first-author papers

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Bio-inspired camera based on flexible optoelectronics



C. Choi et al. Nature Commun. 8, 1664 (2017), C. Choi et al. Nature Electron. 3, 546 (2020)

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Part 2 Synaptic optoelectronic devices for in-sensor preprocessing Integrated SPTr-QLED Deep neural network Joisy optical inputs Optical input Electrical output Short-term plasticity _ Output ii) Long-term potentiatio Neural networ Mu Output Inpu Optic nerves Con Conv Incident image Noisv imag Action potentia In-senso Visual cortex preprocessing SoN on-device Plano-convex lens Preprocessed image preprocessing

C. Choi *et al. Nature Commun.* **11**, 5734 (2020), C. Choi *et al. Sci. Adv.* **8**, *eabq3101* (2022) C. Choi *et al. Adv Sci.* 2304039 (2023), C. Choi *et al. InfoMat.* (*early view*)



Part 1

Bio-inspired camera based on flexible optoelectronics



C. Choi et al. Nature Commun. 8, 1664 (2017), C. Choi et al. Nature Electron. 3, 546 (2020)

Issue of conventional Cameras

1. "Optical aberration" in conventional cameras with single lens optics



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2. "Complicated optics" are expensive, heavy, and bulky.



Inspiration from camera-type eyes



Optical advantages of camera-type eye



⇒ Compared to complicated multi-lens configuration in conventional camera, the camera-type eyes in nature enable the aberration-free imaging using single-lens optics by matching the retina with the curved retina.

Human eye-inspired camera

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Human eye-inspired high-density curved image sensor array



- \Rightarrow Ultrathin image sensor array using MoS₂/graphene heterostructure.
- \Rightarrow High-density curved image sensor array without mechanical failure.
- \Rightarrow Human-eye-inspired configuration (Curved imager and single lens optics).
- \Rightarrow Various optical advantages (reduced optical aberration, single-lens optics).

Aquatic vision in nature



Aquatic vision in nature: Simple optics & Wide FoV



- \Rightarrow Aquatic vision (e.g., fisheye) enables wide field-of-view imaging (>160°) based on a single lens optics, because of their spherical lens and hemispherical retina.
- ⇒ Aquatic vision-inspired camera using silicon hemispherical image sensor array for wide field-of-view, deep depth-of-field imaging with a miniaturized form factor.



Part 2

Synaptic optoelectronic devices for in-sensor preprocessing



C. Choi *et al. Nature Commun.* **11**, 5734 (2020), C. Choi *et al. Sci. Adv.* **8**, *eabq3101* (2022) C. Choi *et al. Adv Sci.* 2304039 (2023), C. Choi *et al. InfoMat.* (*In press*)

Conventional image recognition

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Conventional imaging and data processing systems



Conventional imaging system with conventional processor Conventional imaging system with neuromorphic chips



C. Choi et al. Nature Commun. 11, 5934 (2020)

Conventional Si image sensors – fast photoresponse





C. Choi et al. Nature Electron. 3, 546 (2020)



Curved neuromorphic image sensor



cNISA inspired by human visual recognition system



- ⇒ Curved neuromorphic image sensor array for aberration-free imaging with a single lens optics and preprocessing with photon-triggered synaptic plasticity.
- \Rightarrow The preprocessed image, of which noise is reduced and contrast is enhanced, can be derived from the noisy optical inputs through a single-readout operation.

Synaptic photodetector for CP light



Synaptic photodetector for direct CP light detection



- ⇒ Synaptic photodetector for direct CP light detection was developed by depositing chiral organogelators onto IGZO phototransistor.
- ⇒ DPPPT-cPTr shows synaptic photoresponse, generating the accumulated photocurrent in response to the frequent CP light of target orientation only.

All-or-None potentiation of synapse



Integrated device inspired by all-or-none potentiation of human synapse



All-or-None on-device preprocessing

- ⇒ All-or-none potentiation as well as synaptic plasticity realized by integrating the synaptic phototransistors and quantum-dot light-emitting diodes (QLED).
- ⇒ SPTr with photon-triggered synaptic plasticity & QLED with threshold switching characteristics enable "Signal-or-None (SoN) on-device preprocessing".

Improvement in image recognition



Improvement in image recognition based on deep neural network

$I_{OLED} = \alpha_{OLED} (\exp(P_{UV} / \tau_{OLED}) - 1)$ Noisy images $I_{SPTr} = \alpha_{SPTr} P_{UV}$ 40 1.0 30 I_{SPTr} (nA) l_{aled} (mA) 20 0.5 10 In-sensor preprocessed images (simulation) 0 0 10 20 30 2 3 40 1 n Light intensity (µW cm⁻²) I_{SPTr} (nA) Noisy images 100 > 89% > 86% In-sensor SoN on-device preprocessed images (simulation) Accuracy (%) Accuracy (%) 100 SoN on-device < 50% 50 < 36% 50 20 Control Noisy In-sensor Sol 15 Epochs

- ⇒ Simulated images (preprocessed images and SoN preprocessed images) were prepared from noisy images using the empirically fitted parameters.
- ⇒ Recognition rate of such images were evaluated using ResNet50 trained with standard training dataset.
- ⇒ SoN preprocessing significantly improve the image recognition because of significant reduction of background noise.

C. Choi et al. Sci. Adv. 8, eabq3101 (2022)

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Group Members



Collaborators

Prof. Dae-Hyeong Kim (SNU) Prof. Donhee Ham (Harvard) Dr. Jung Ah Lim (KIST) Prof. SungWoo Nam (UC Irvine) Prof. Young Min Song (GIST) Prof. Taeghwan Hyeon (SNU) Prof. Moon Kee Choi (UNIST) Dr. Do Kyung Hwang (KIST)



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Fhank you

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