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Reporting Summary

Nature Portfolio wishes to improve the reproducibility of the work that we publish. This form provides structure for consistency and transparency in reporting. For further information on Nature Portfolio policies, see our <u>Editorial Policies</u> and the <u>Editorial Policy Checklist</u>.

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FOI :	an statistical analyses, commit that the following items are present in the figure regend, than lext, or internous section.
n/a	Confirmed
	\square The exact sample size (n) for each experimental group/condition, given as a discrete number and unit of measurement
	A statement on whether measurements were taken from distinct samples or whether the same sample was measured repeatedly
	The statistical test(s) used AND whether they are one- or two-sided Only common tests should be described solely by name; describe more complex techniques in the Methods section.
\boxtimes	A description of all covariates tested
\boxtimes	A description of any assumptions or corrections, such as tests of normality and adjustment for multiple comparisons
	A full description of the statistical parameters including central tendency (e.g. means) or other basic estimates (e.g. regression coefficient) AND variation (e.g. standard deviation) or associated estimates of uncertainty (e.g. confidence intervals)
	For null hypothesis testing, the test statistic (e.g. <i>F</i> , <i>t</i> , <i>r</i>) with confidence intervals, effect sizes, degrees of freedom and <i>P</i> value noted <i>Give P values as exact values whenever suitable.</i>
\boxtimes	For Bayesian analysis, information on the choice of priors and Markov chain Monte Carlo settings
\boxtimes	For hierarchical and complex designs, identification of the appropriate level for tests and full reporting of outcomes
\boxtimes	Estimates of effect sizes (e.g. Cohen's <i>d</i> , Pearson's <i>r</i>), indicating how they were calculated
	Our web collection on <u>statistics for biologists</u> contains articles on many of the points above.

Software and code

Policy information about availability of computer code

Data collection

No software was used for data collection.

Data analysis

We trained our UIOS and other comparison methods on the public platform Pytorch and Nvidia Geforce RTX 3090 GPU (24G). Adam was adopted as the optimizer to optimize the model. The network implementation used the PyTorch deep learning framework (PyTorch 1.7.1). Result analyses are performed with custom code written in Python (python 3.9.13) with several packages, including numpy (numpy 1.21.5), and sklearn (sklearn 1.0.2). All codes are available at (https://github.com/LooKing9218/UIOS)

For manuscripts utilizing custom algorithms or software that are central to the research but not yet described in published literature, software must be made available to editors and reviewers. We strongly encourage code deposition in a community repository (e.g. GitHub). See the Nature Portfolio guidelines for submitting code & software for further information.

Data

Policy information about availability of data

All manuscripts must include a <u>data availability statement</u>. This statement should provide the following information, where applicable:

- Accession codes, unique identifiers, or web links for publicly available datasets
- A description of any restrictions on data availability
- For clinical datasets or third party data, please ensure that the statement adheres to our <u>policy</u>

Data from JSIEC1000 is available at (https://www.kaggle.com/datasets/linchundan/fundusimage1000).

Data from VOC2012 is available Additional data sets supporting	e at (https://retouch.grand-challenge.org). e at (https://host.robots.ox.ac.uk/pascal/VOC/voc2012). g the findings of this study were not publicly available due to the confidential policy of National Health Commission of China, but are ng authors upon reasonable request.
Research involving	g human participants, their data, or biological material
Policy information about stu and sexual orientation and r	idies with <u>human participants or human data</u> . See also policy information about <u>sex, gender (identity/presentation),</u> ace, ethnicity and racism.
Reporting on sex and gend	der N/A
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Population characteristics	N/A
Recruitment	N/A
Ethics oversight	N/A
Note that full information on th	e approval of the study protocol must also be provided in the manuscript.
	that is the best fit for your research. If you are not sure, read the appropriate sections before making your selection.
Life sciences	Behavioural & social sciences
Life sciences	study design
All studies must disclose on	these points even when the disclosure is negative.
datasets were use photo pu with 17, tomogra for traini	cted 10,034 fundus images of 8 different fundus diseases or normal condition. Two non-target categories retinal diseases (NTC dataset with 1,380 samples and NTC-JSIEC dataset with 502 samples) and one low quality image dataset with 1,066 samples ed to investigate the capability of UIOS to detect fundus abnormalities outside the categories of the training set. Three non-fundus ublic datasets were used to evaluate the performance of AI models in detecting OOD samples. The first was the VOC2012 dataset, 125 natural images of 21 categories. The second was RETOUCH dataset which consisted of 6,936 2-D retinal optical coherence phy images. The third was our OCTA dataset collected from our eye clinic, consisting of 304 2D retinal OCTA images. The sample size ng, validation and test was informed by the existing literature and works in the field of deep learning on fundus retinal images and nty study(Dolezal et al., 2022; Engelmann et al., 2022).
fundus c specialis (Supplen	collection and labelling procedure are shown in Supplementary Fig. 4. Fundus images from 5 eye clinics with different models of ameras were collected. Two trained graders performed the annotation independently. If their results were inconsistent, a retinal subtwith more than 10 years experience would make the final decision. Those images do not belong to any criteria mentioned in nentary table.16) were considered unclassifiable and excluded. Moreover, we removed the images in the categories of massive hard cotton-wool spots, preretinal hemorrhage, fibrosis and laser spots from JSIEC to avoid confusion as reviewer suggested.
given as ensembl and Nvic were set the imag	e used for training the deep-learning models are made publicly available for the reproducibility purpose. Statistical analysis has been well. Specifically, we trained our UIOS and other comparison methods including standard AI model, Monte-Carlo drop-out (MC-Drop), e models, time-augmentations (TTA), using entropy across the categorical class probabilities (Entropy), on the public platform Pytorch lia Geforce RTX 3090 GPU (24G). Adam was adopted as the optimizer to optimize the model. Its initial learning rate and weight decay to 0.0001 and 0.0001, respectively. The batch size was set to 64. To improve the computational efficiency of the model, we resized e to 256×256.Meanwhile, online random left-right flipping was applied for data augmentation. All code is available at (https://om/LooKing9218/UIOS) for replication of all experimental results.

Blinding Image capture was based on daily practice of multi-centers which did not require blinding. Annotators for image labelling were blinded to the ground truth and were not involved in image collection. Engineers did not inspect the contents of datasets in algorithm developing and were blinded to the effect evaluation.

dataset. These images were randomly divided into training (6,016), validation (2,008) and test sets (2,010) in the ratio of 6:2:2.

We collected 10,034 fundus images of 8 different fundus diseases or normal condition. They were named the primary target-categories (TC)

Reporting for specific materials, systems and methods

Randomization

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system or method listed is relevant to your study. If you are	not sure if a list item applies to your research, read the appropriate section before selecting a response.
Materials & experimental systems	Methods
n/a Involved in the study	n/a Involved in the study

We require information from authors about some types of materials, experimental systems and methods used in many studies. Here, indicate whether each material,

Triateriais & experimental systems				
n/a	Involved in the study	n/a	Involved in the study	
\times	Antibodies	\boxtimes	ChIP-seq	
\boxtimes	Eukaryotic cell lines	\boxtimes	Flow cytometry	
\boxtimes	Palaeontology and archaeology	\times	MRI-based neuroimaging	
\boxtimes	Animals and other organisms			
\times	Clinical data			
\boxtimes	Dual use research of concern			
\boxtimes	Plants			