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香港城市大學

City University of Hong Kong

Automated Non-invasive Analysis of Motile Sperms Using Cross-scale Guidance Network

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PhD Student

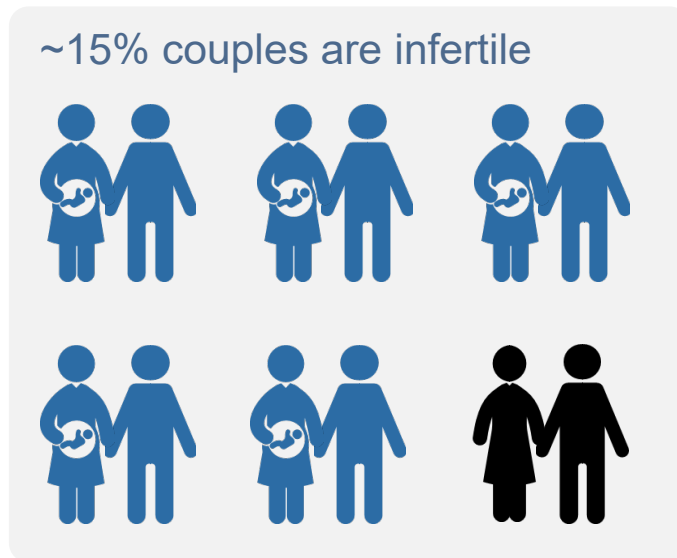
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May 16, 2024



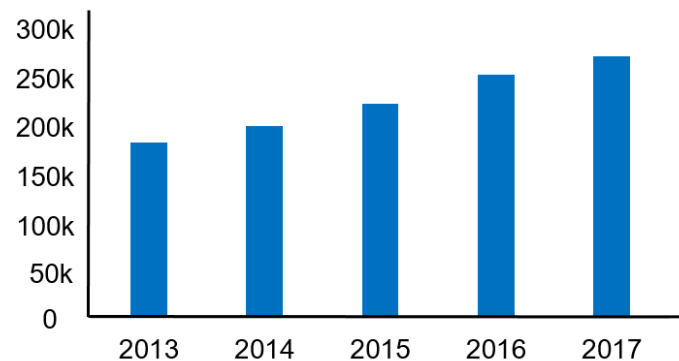
Multiscale Automation
and Robotics Laboratory

In vitro fertilization (IVF)



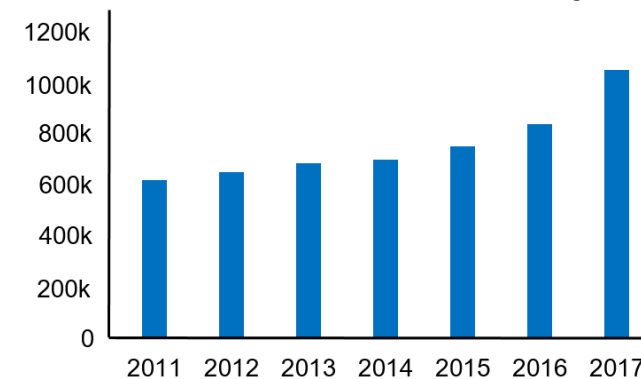
- ~113 million infertile population
Lancet, 2016
- >8 million IVF children born
ICMART, 2018
- \$25 billion market – as 2019
- \$41 billion market – by 2026
www.economist.com/business/2019/08/08

US treatment cycles



CDC National Summary Report

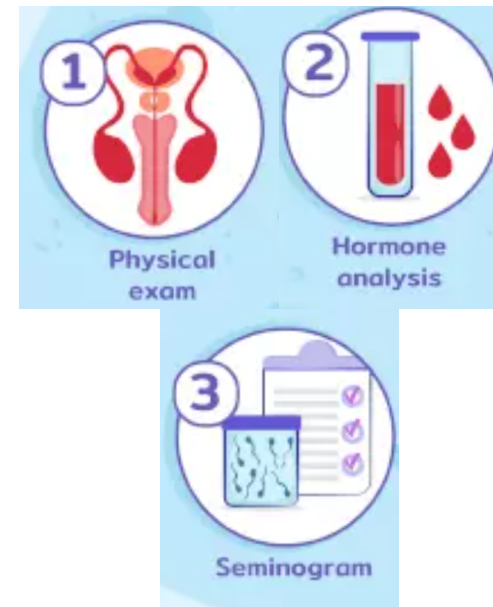
China treatment cycles



Forward Institute Summary Report

Male Infertility

- Male fertility problems contribute to 30% of infertility cases (You et al. 2021).
- The morphology and motility of sperm are critical for male fertility.



Manual Inspection

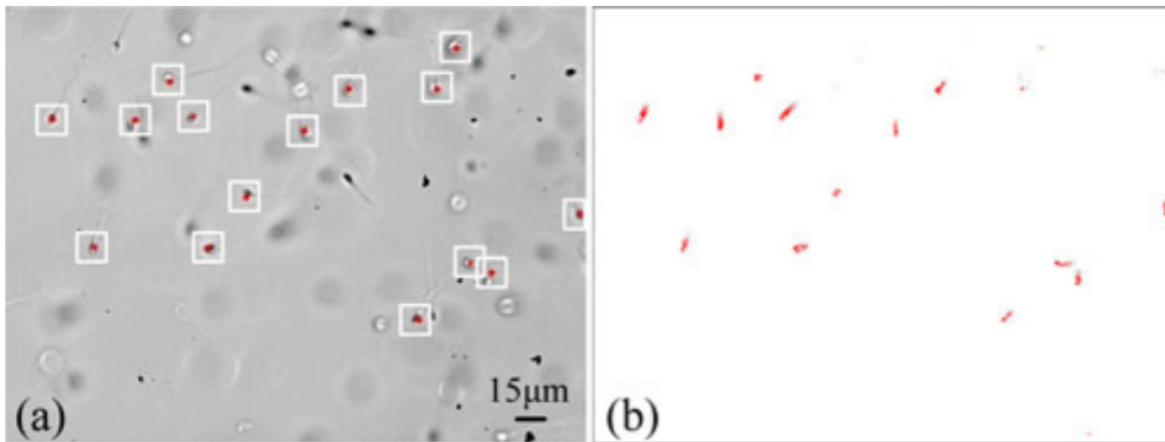
- Manual inspection and selection are laborious.



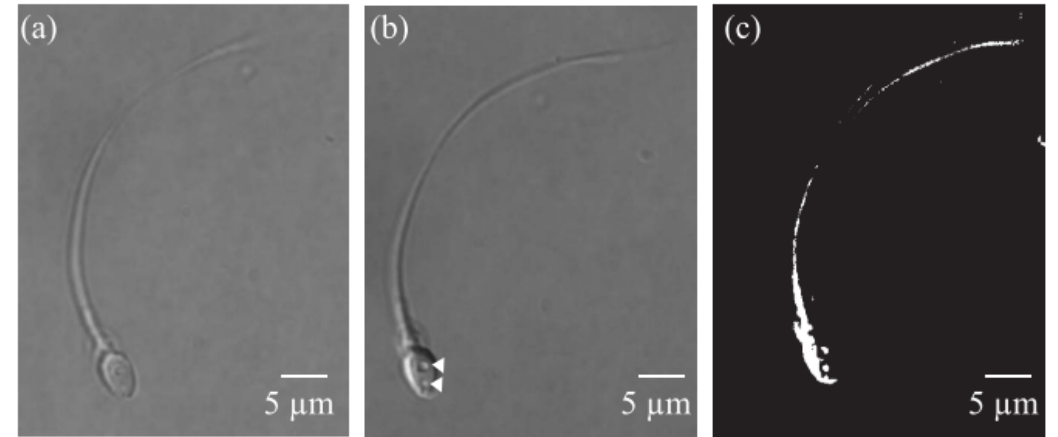
Automated sperm analysis

Conventional computer vision:

- Kalman filter, track sperm head [1]
- Differential interference contrast (DIC), identify sperm morphology [2]



(Liu et al., TBME, 2012).

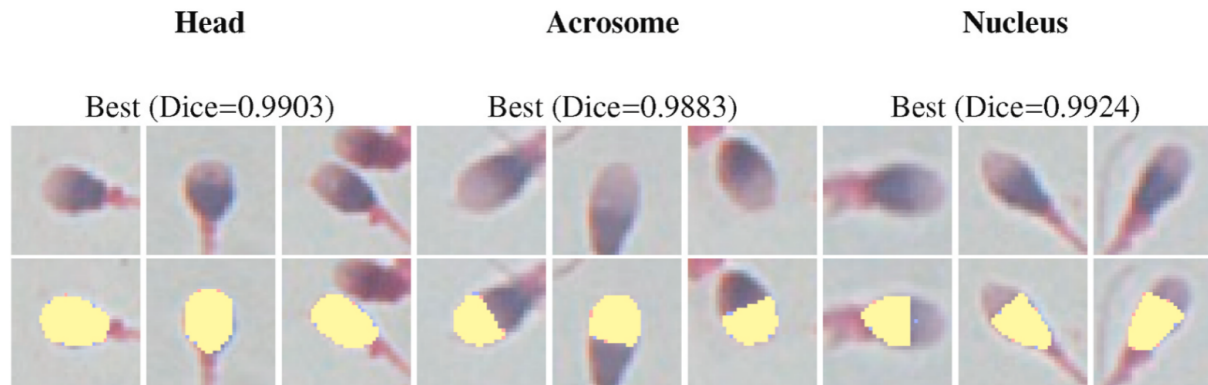


(Dai et al., TMI, 2018).

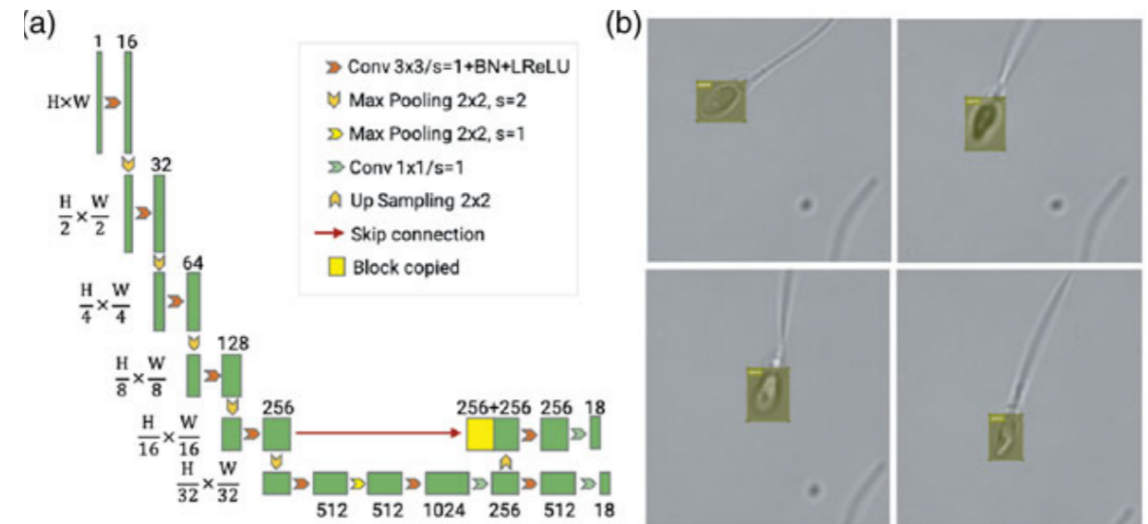
Automated sperm analysis

Machine-learning-based computer vision:

- UNet , sperm head segmentation.
- YOLO, track sperm head.



(Marín et al., Computers in Biology and Medicine, 2021)



(Liu et al., Microscopy and Microanalysis, 2022)

Problems and Challenges

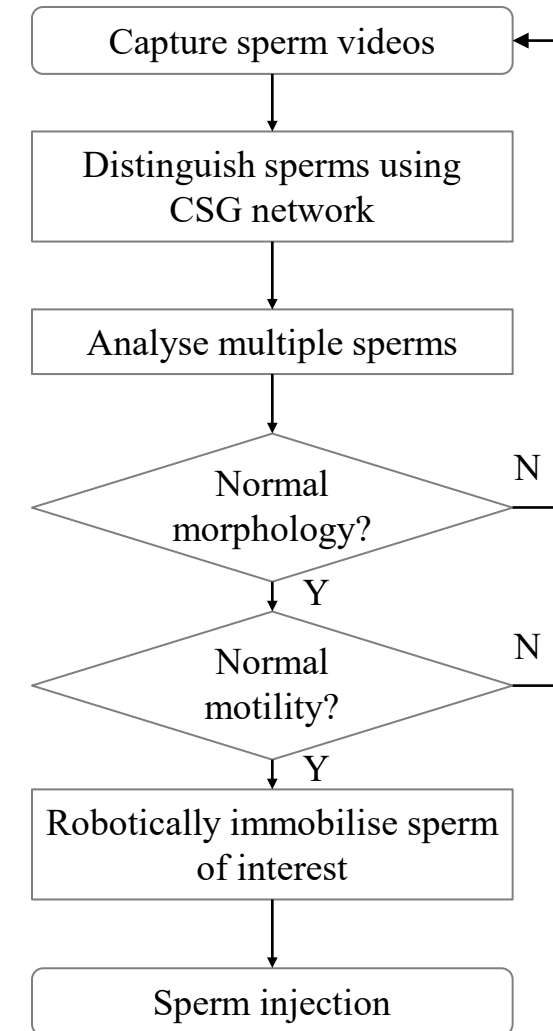
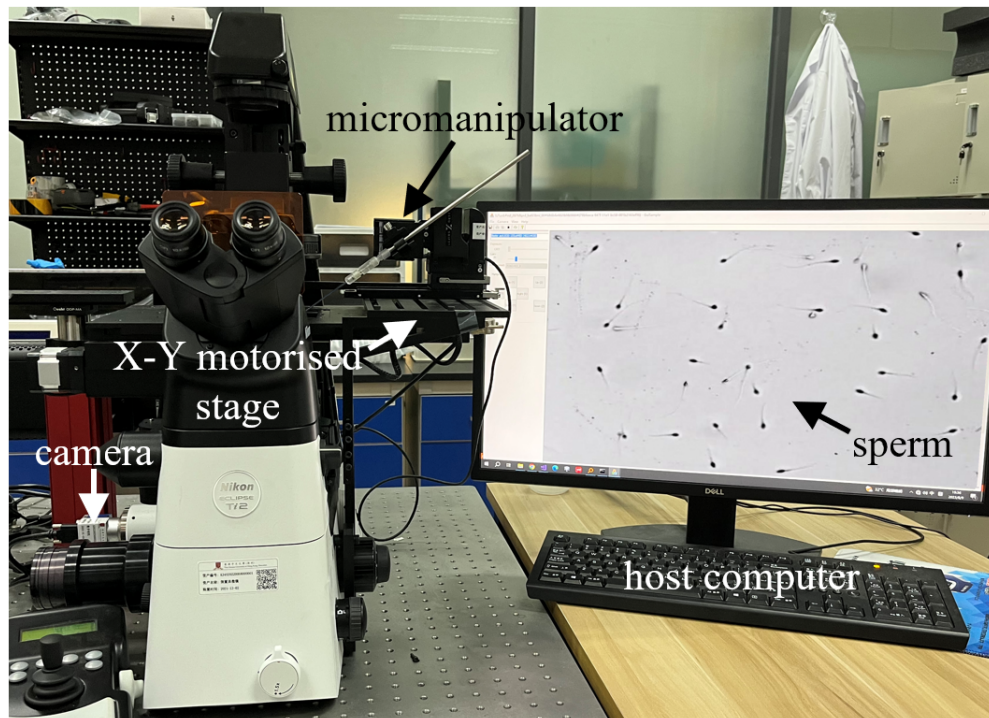
- Morphology and motility are not analyzed simultaneously.
- Averaged sperms per image are less as magnification increases.
- Dyes and fluorescences make sperms clinically unavailable.
- Too small to be detected. Less than 1% area ratio of a petri dish under 20× objective lens.

Main Contributions

- Introduce a novel architecture that alleviate compression artifacts.
- Measure sperm's morphology and motility simultaneously.
- Analyze sperm in a non-invasive manner at 20x objectives.

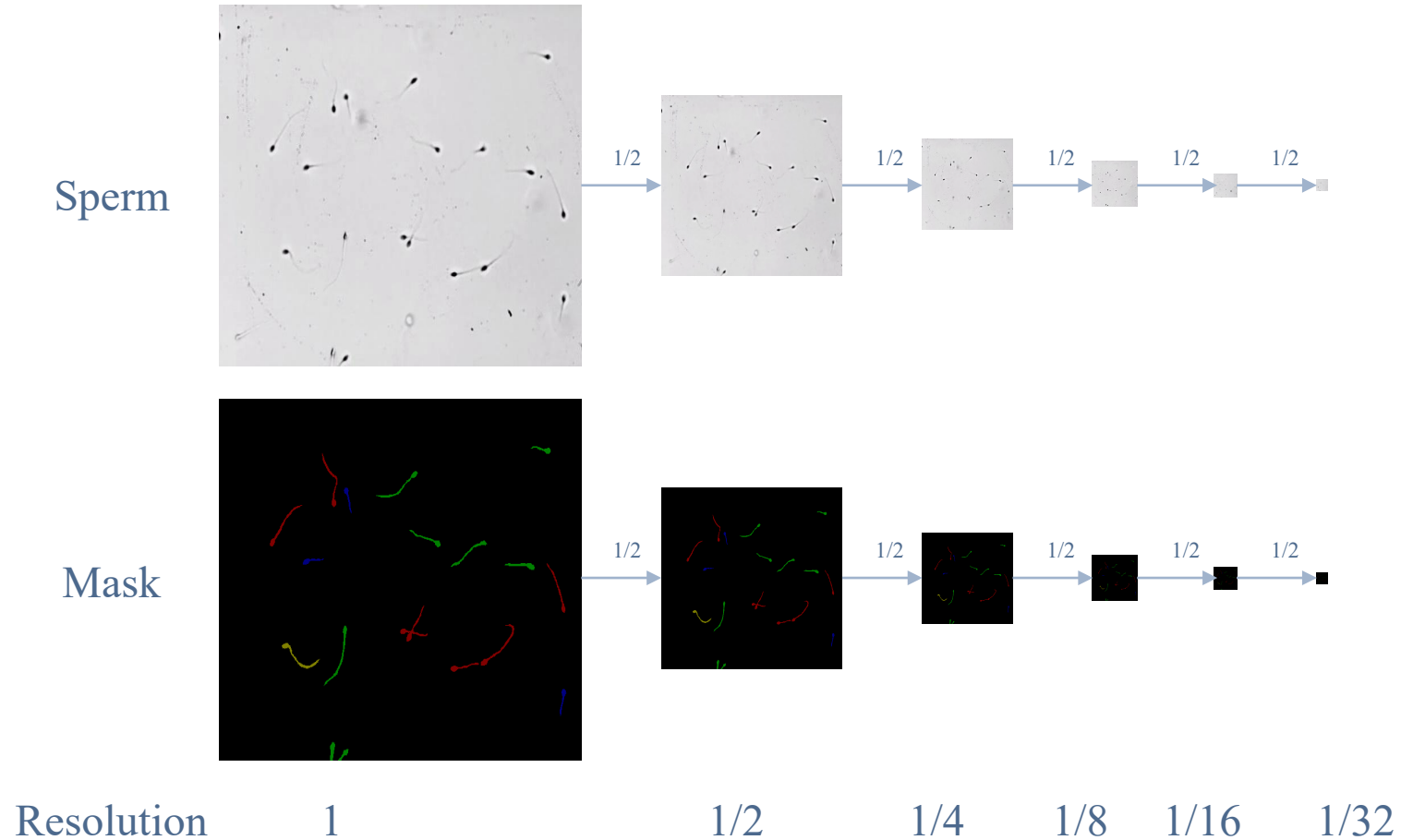
Micro-robotic cell manipulation System

Nikon microscope with 20× objective lens, CMOS camera, 3-DOF micromanipulator (MP-285).



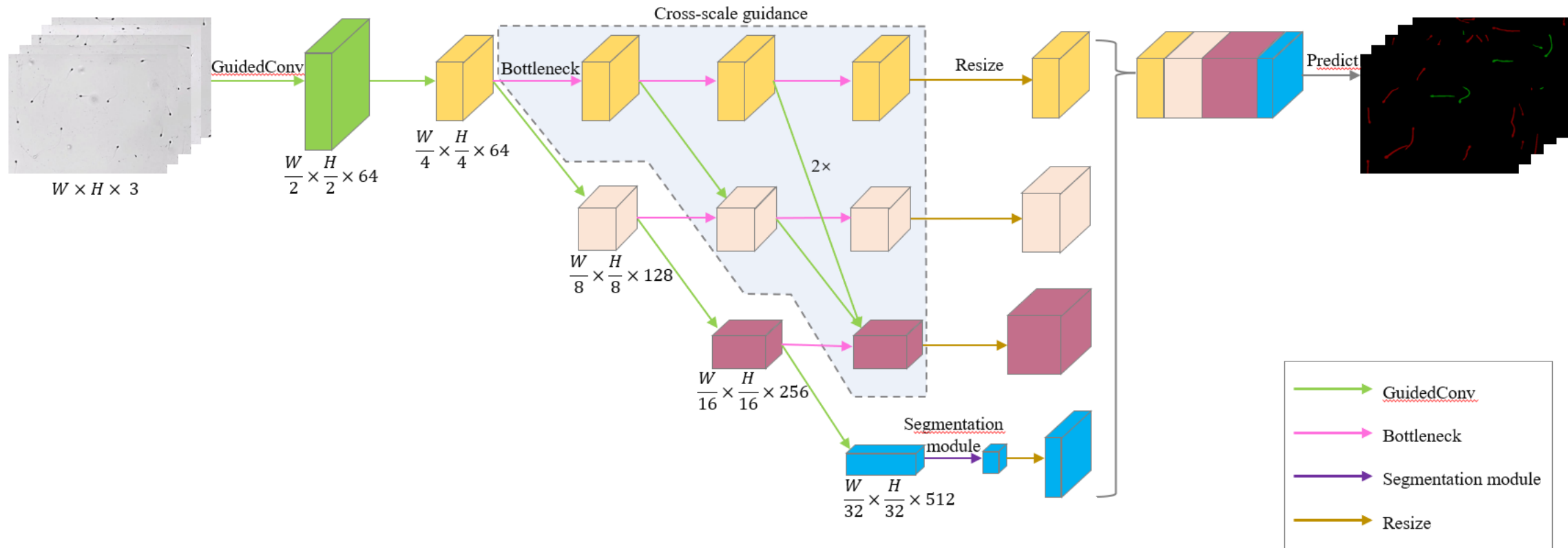
Motivation

Information loss during processing



Cross-scale Guidance Network (CSG Network)

Main component: cross-scale feature map guide

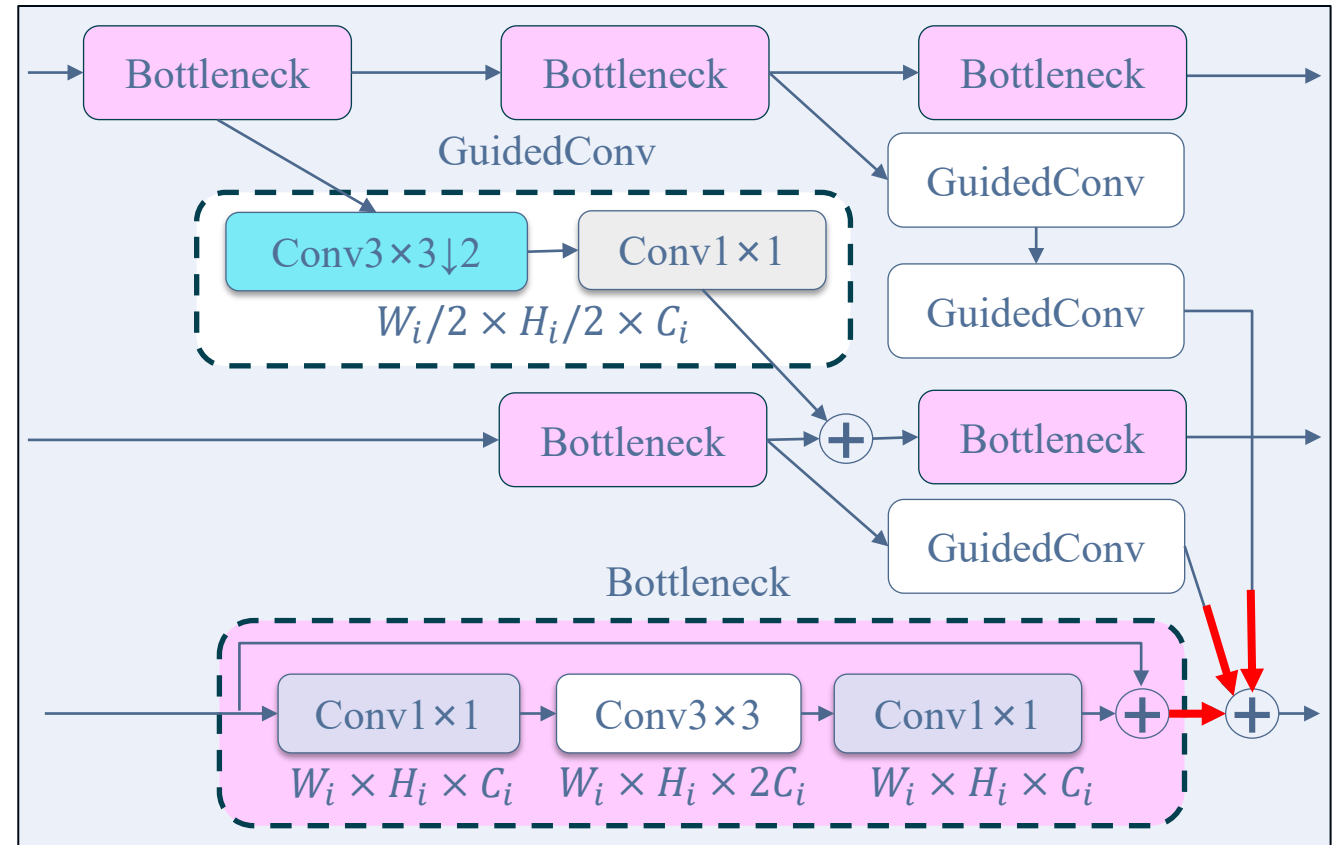


Cross-scale feature map guide

R_x : input tensor

$f_{xr}(R_x)$: (r-x) 3×3 strided convolution

$$R' = f_{1r}(R_1) + f_{2r}(R_2) + f_{3r}(R_3)$$

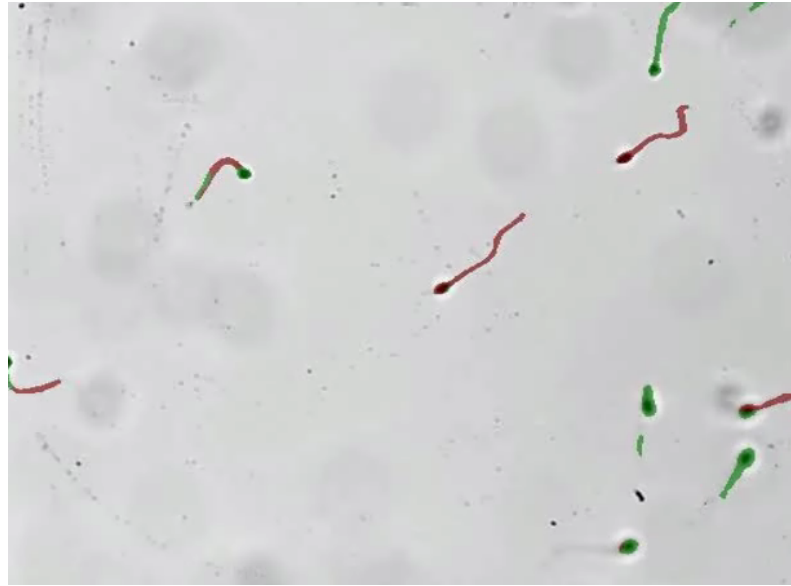


Video Outputs

Speed $\times 1.5$



Coarse



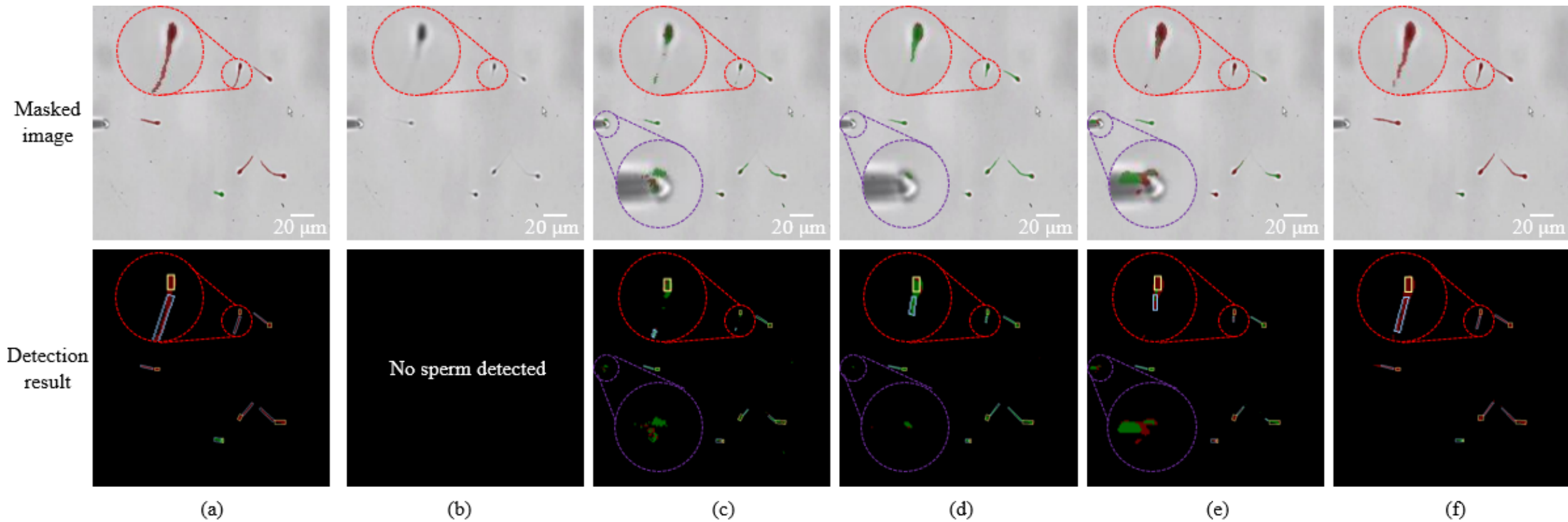
Medium



Dense

Error Analysis

Visualisation of segmentation ground truth (a) and segmentation results using (b) ResNet50 + DeepLabV3, (c) SegNet, (d) UNet, (e) UNet++, and (f) **CSG Network + DeepLabV3**.



Segmentation quantity results

- Achieved highest mIoU of 51.89%.
- Exceeding 21% and 32% for normal and abnormal sperm segmentation

Segmentation IoU and mIoU (Unit:%) for various methods.
 $\text{IoU} = \text{TP} / (\text{TP} + \text{FP} + \text{FN})$.

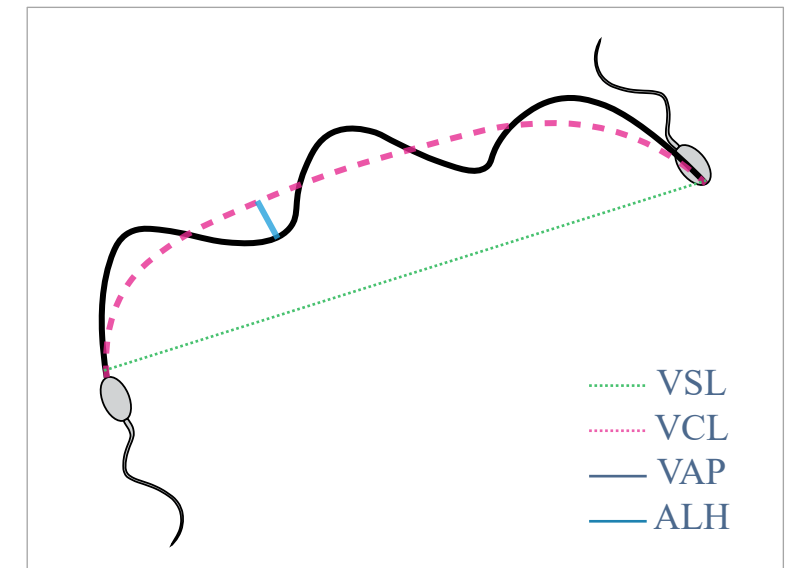
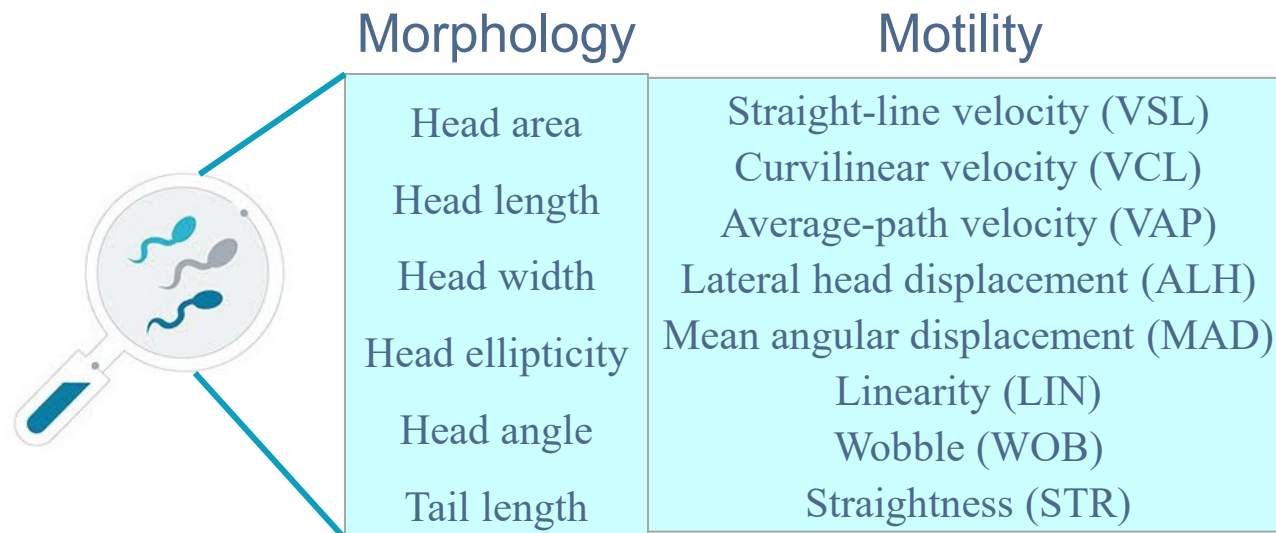
Method	Module	Background	Normal	Abnormal	mIoU
SegNet	-	99.11	7.11	24.93	42.41
UNet	-	99.30	13.82	34.06	48.25
UNet++	-	99.29	18.29	33.29	48.40
ResNet50	OCR	98.86	0.00	0.00	33.98
	LR-ASPP	98.87	1.36	5.14	35.19
	DeepLabV3	98.86	0.00	0.02	33.27
CSG Network (ours)	OCR	99.31	21.23	32.77	51.45
	LR-ASPP	99.30	22.41	33.36	51.64
	DeepLabV3	99.31	21.61	34.60	51.89

Case Studies

Sperm No.3 is the only healthy sperm

TABLE II: AUTOMATED QUANTIFICATION OF FIVE SPERM SAMPLES (AU: ARBITRARY UNIT).

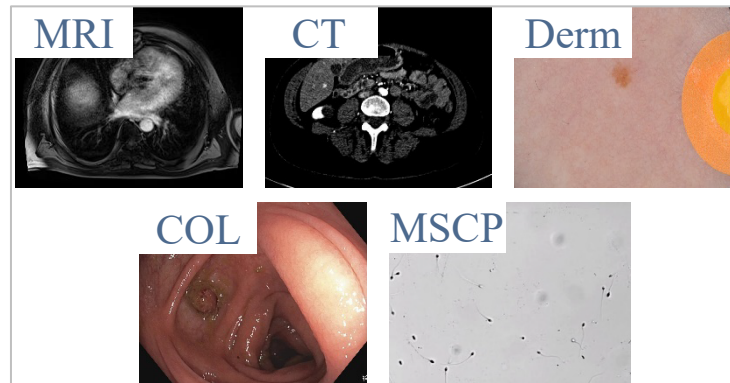
Sperm No.	Morphology							Motility									Healthy
	area (μm^2)	length (μm)	width (μm)	ellipticity (AU)	angle ($^\circ$)	tail length (μm)	Normal	VSL ($\mu\text{m/s}$)	VCL ($\mu\text{m/s}$)	VAP ($\mu\text{m/s}$)	ALH ($\mu\text{m/s}$)	MAD ($^\circ$)	LIN (AU)	WOB (AU)	STR (AU)	Normal	
1	14.50	5.62	3.00	1.88	90.00	15.74	✗	12.64	12.64	12.65	0.66	0.62	1.00	1.00	1.00	✓	✗
2	14.50	5.00	3.00	1.67	0.00	38.93	✓	0.68	0.80	0.98	0.82	0.97	0.85	1.22	0.70	✗	✗
3	17.50	5.74	3.50	1.64	2.73	40.45	✓	11.97	12.01	12.01	0.99	0.22	1.00	1.00	1.00	✓	✓
4	19.56	6.25	2.00	3.13	72.25	33.83	✗	0.20	0.20	0.21	0.29	0.47	0.98	1.01	0.98	✗	✗
5	25.31	4.75	3.75	1.27	0.00	33.03	✓	0.76	0.76	0.76	0.10	3.19	1.00	1.00	1.00	✗	✗



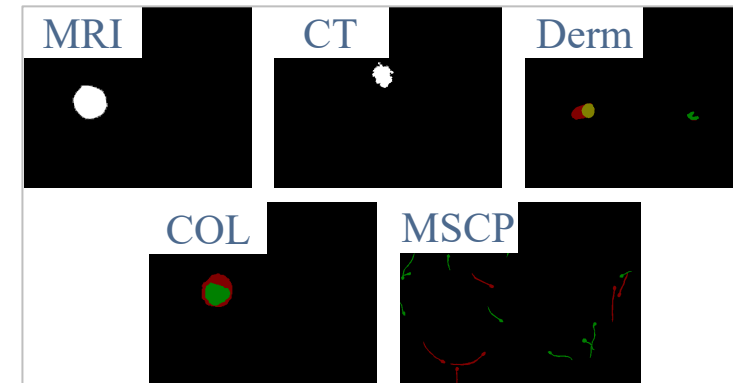
Conclusion

- Outperformed other SOTA methods by over **3.59%** mIoU.
- Selected the healthy sperm among samples non-invasively.

General small medical object detection



Images of
Small Medical Objects



Segmentation
Results

Acknowledgment



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