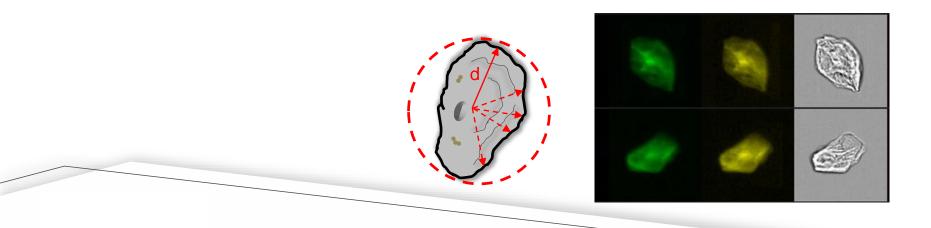
Developmental validation of a novel approach for determining time-since-deposition of trace DNA evidence

Christopher J. Ehrhardt

Department of Forensic Science, Virginia Commonwealth University



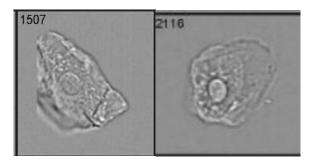
Time Since Deposition (TSD)

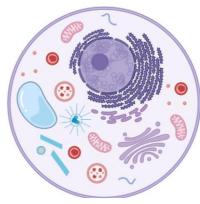
- One of the primary questions around DNA evidence is <u>when</u> it was deposited
- Various methods for TSD have been proposed over the years.... none have focused on 'touch' biological evidence



• Epidermal cell populations are ... different !

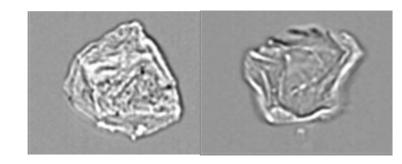
Stratified epithelial cells (saliva)

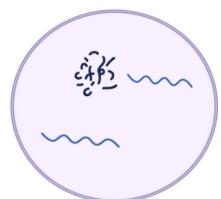




- Nucleus
- Organelles
- Vesicles
- Mitochondria
- Cytoskeleton
- Protein aggregations

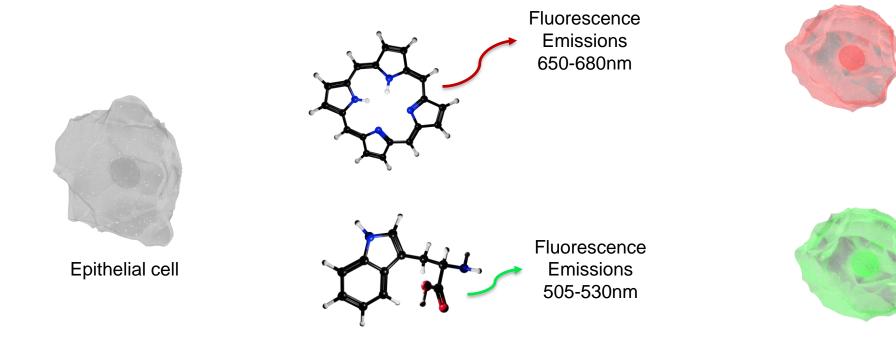
Shed epidermal cells, corneocytes



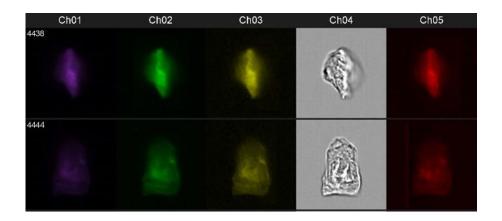


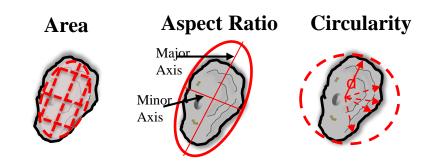
- Lipoproteins
- Keratins
- Degraded proteins, DNA

- Autofluorescence profiling: snapshot of cellular biochemistry,
- Compounds fluoresce at different wavelengths, intensity varies with compound abundance within cell



- Measure wavelengths, intensity of autofluorescence of cells...
- Combine autofluorescence with morphological measurements
- Fast and nondestructive!





 Previous work used autofluorescence to differentiate tissue sources & identify contributor cell populations in touch DNA samples

PLOS ONE

RESEARCH ARTICLE

Rapid differentiation of epithelial cell types in aged biological samples using autofluorescence and morphological signatures

Emily R. Brocato, M. Katherine Philpott, Catherine C. Connon, Christopher J. Ehrhardt*



Contents lists available at ScienceDirect

Forensic Science International: Genetics

journal homepage: www.elsevier.com/locate/fsigen

Differentiation of vaginal cells from epidermal cells using morphological and autofluorescence properties: Implications for sexual assault casework involving digital penetration

Sarah Ingram ^a, Arianna DeCorte ^a, Amanda Elswick Gentry ^b, M. Katherine Philpott ^a, Taylor Moldenhauer ^a, Sonja Stadler ^c, Cory Steinberg ^c, Jonathan Millman ^d, Christopher J. Ehrhardt ^{a,*}

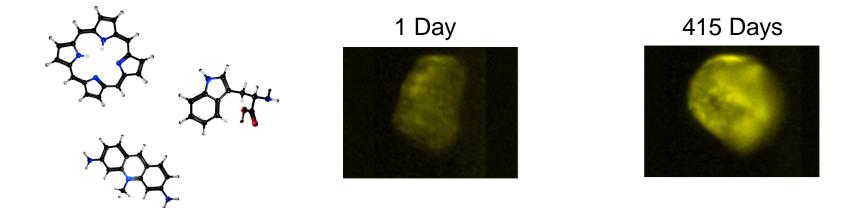
F1000Research	F1000Research 2016, 5:180 Last updated: 16 FEB 2016
RESEARCH NOTE	CrossMark ¢ click for updates
Analysis of red autofluorescence	e (650-670nm) in epidermal cell
populations and its potential for	distinguishing contributors to
'touch' biological samples [version of the second sec	on 1; referees: awaiting peer
review]	
Cristina E. Stanciu, M. Katherine Philpott, Ec Christopher J. Ehrhardt	duardo E. Bustamante, Ye Jin Kwon,
Department of Forensic Science, Virginia Commonwealth University, Ri	ichmond, VA, USA
DOI 10.1007/s00216-017-0364-0	
RESEARCH PAPER	

Analysis of cellular autofluorescence in touch samples by flow cytometry: implications for front end separation of trace mixture evidence

M. Katherine Philpott¹ • Cristina E. Stanciu¹ • Ye Jin Kwon¹ • Eduardo E. Bustamante¹ • Susan A. Greenspoon² • Christopher J. Ehrhardt¹

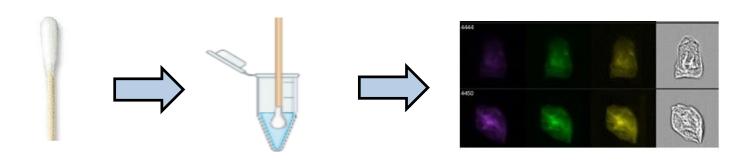
• Even though biochemistry of touch epidermal cells are unusual, there are dozens of compounds that can autofluoresce

• As they degrade, autofluorescence should change with time !



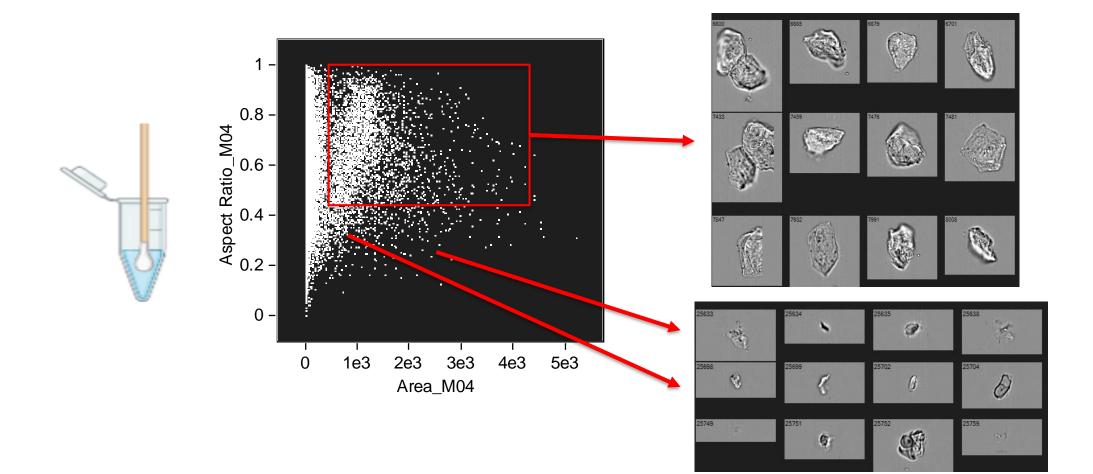
TSD Signature Development

- Created time series of 'touch' biological samples:
 - >80 different contributor cell populations from ~50 individuals
 - Aged between 1 day and ~2 yrs, various substrates
 - Swabs eluted directly in water, cells analyzed with flow cytometry, no other steps or reagents!



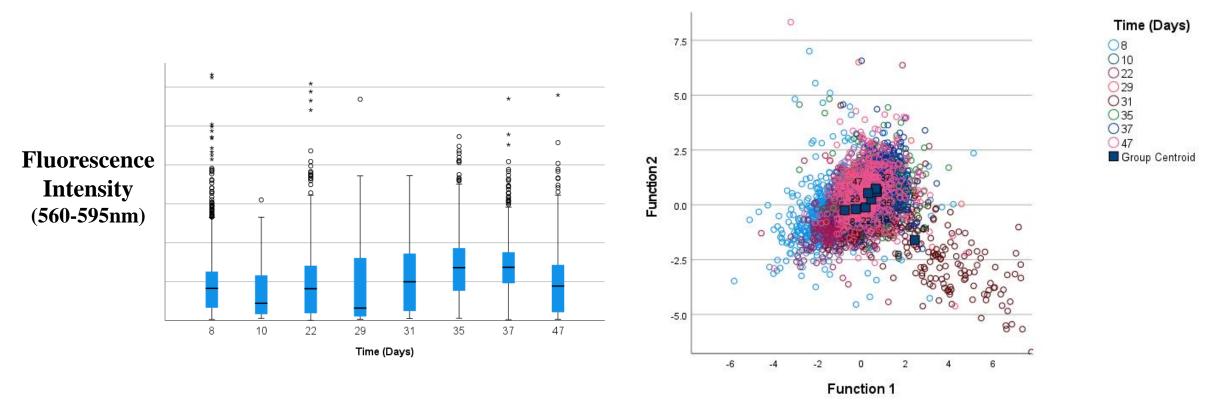
Phase I: Results

• Touch epidermal cell populations: what do we analyze?



Phase I: Results

• Complex trends across entire time series....



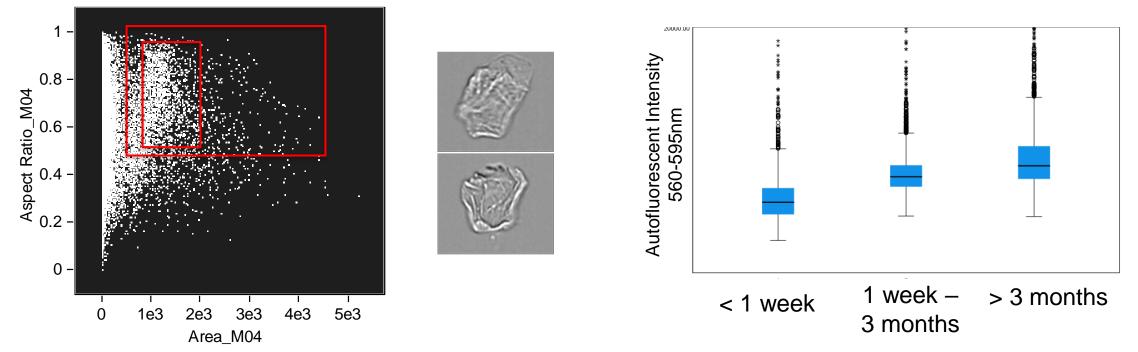
• Complete resolution of time points not likely, need a new strategy

• Is TSD to a specific day necessary for forensics?



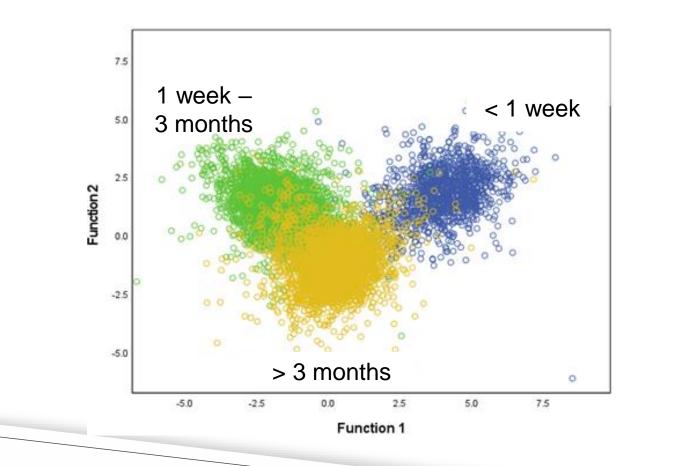
Phase I: Results

• Two changes: (1) modeled autofluorescence across discrete time intervals, (2) narrowed the subpopulation of cells,



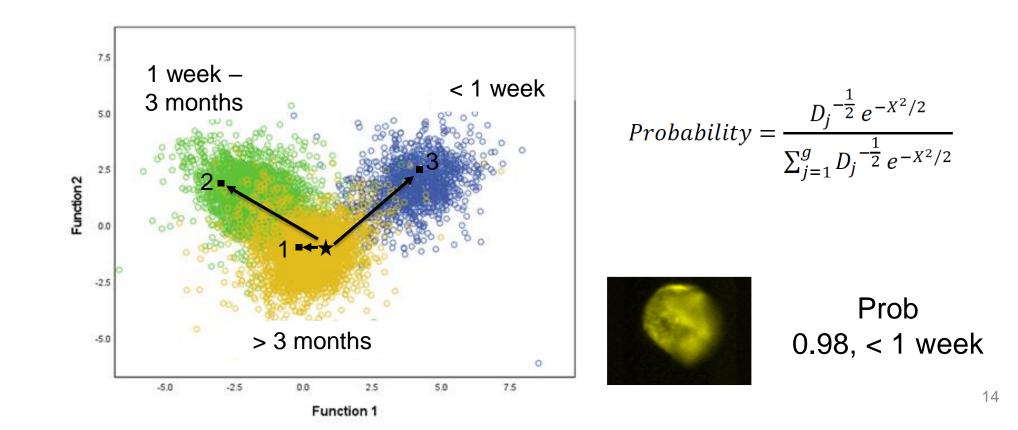
Phase I: Results

• Multivariate modeling of time intervals (Linear Discriminant Analysis)



Wilks lambda =0.18, p<0.001

• Estimate TSD of individual cells within sample based on multivariate distances, also calculate posterior probability of tissue ID



• Tabulate TSD and probability for *every* cell in a given sample:

	Cell	TSD	Post Prob
	 - ▶ 1	<1 week	0.98
	▶ 2	< 1 week	0.95
	. 3	< 1 week	0.78
>	4	< 1 week	0.99
	5	< 1 week	0.90
	6	~ 2 months	0.56
	7	< 1 week	0.82
	300	< 1 week	0.96

• For one cell population, metrics can be combined

Contributor ID	Total Cells	Time Since Deposition (Actual)	Time Since Deposition (Predicted)	Posterior Probability (Average)
B32	93	22 days	1week-3months	0.93
J72	192	22 days	1week-3months	0.98
L12	312	22 days	1week-3months	0.99
K47	67	29 days	1week-3months	0.97
H21	271	31 days	1week-3months	0.95
I 66	38	35 days	1week-3months	0.76

- ~80 samples classified into three time intervals: < 1 week, 1 week to 2 months, >3 months
- Correct TSD every time the sample had >75 cells & the average posterior probability >0.90 !
- TSD possible for ~50% of touch samples

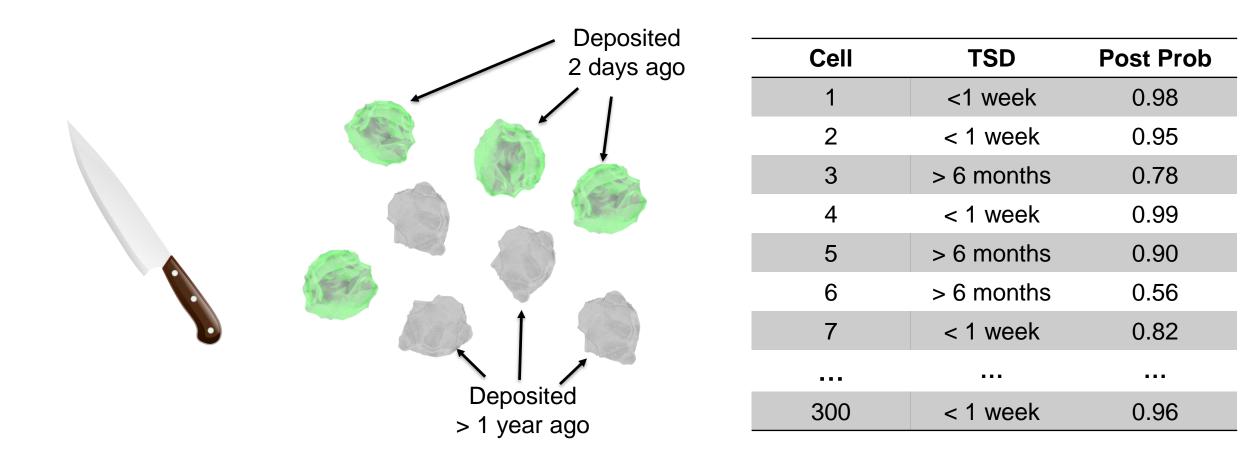
			Highest Group		
Sample	Total	Time/Group	Total	Group	Posterior
ID	Events		Events		Probability
C58	219	1 day / 1	218	1	0.999
L49	39	1 day / 1	36	1	0.999
N90	156	1 day / 1	114	3	0.892
P22	184	1 day / 1	180	1	0.995
D68	3	3 days / 1	2	1	1.000
I 66	15	3 days / 1	15	1	0.999
P22	83	3 days / 1	75	1	0.970
T04	12	3 days / 1	11	1	0.957
C58	331	4 days / 1	329	1	0.998
E10	54	4 days / 1	49	1	0.957
P22	180	4 days / 1	178	1	0.997
B32	27	8 days / 2	20	2	0.855
H71	18	8 days / 2	14	2	0.837
166	76	8 days / 2	69	2	0.953
J72	856	8 days / 2	826	2	0.962
V73	61	8 days / 2	57	2	0.953
H21	56	10 days / 2	51	2	0.964
A24	129	22 days / 2	120	2	0.927
B32	93	22 days / 2	89	2	0.938
J72	192	22 days / 2	188	2	0.984
L12	312	22 days / 2	307	2	0.994
K47	67	29 days / 2	65	2	0.972
H21	271	31 days / 2	232	2	0.954
<mark>166</mark>	38	35 days / 2	18	2	0.759
J72	707	35 days / 2	626	2	0.904
B32	779	37 days / 2	655	2	0.941
L12	49	47 days / 2	26	3	0.821
V73	186	47 days / 2	105	2	0.856
L12	35	59 days / 2	29	3	0.883
B32	12	72 days / 3	10	3	0.985
A24	8	75 days / 3	7	3	0.858

- What about samples with less than <75 cells?
- Tested other frameworks:
 - Generalized Linear Mixed Model (GLM),
 - Gradient Boosting Machine (GBM),
 - Ridge Regression Model (RRM)
- Binary TSD estimates (i.e., less than or more than a week old, ...)

• GLM >99% accuracy with TSD without minimum number of cells

TSD Estimate	Generalized Linear Mixed Model	CSH Spring Harbor Laboratory bioRxiv
< 7 days	0.996	THE PREPRINT SERVER FOR BIOLOGY
< 30 days	0.913	
< 60 days	0.919	New Results A Follow this preprint Comparison of three quantitative approaches for estimating time-since-
< 90 days	0.847	deposition from autofluorescence and morphological profiles of cell populations from forensic biological samples
< 120 days	0.891	Amanda Elswick Gentry, Sarah Ingram, M. Katherine Philpott, Kellie J. Archer, Christopher J. Ehrhardt
< 180 days	0.961	doi: https://doi.org/10.1101/2023.04.19.537512

• What about mixture samples with multiple TSDs ?



• Question for mixture samples: "Are there fresh cells present ?"



			Cell Counts	
Substrate	TSD	< 1 week	1 week-2 mos	> 3 months
Door Knob 2	~ 1 year			1
Door Knob 3	~ 1 year			
Airhood 1	~ 1 year			

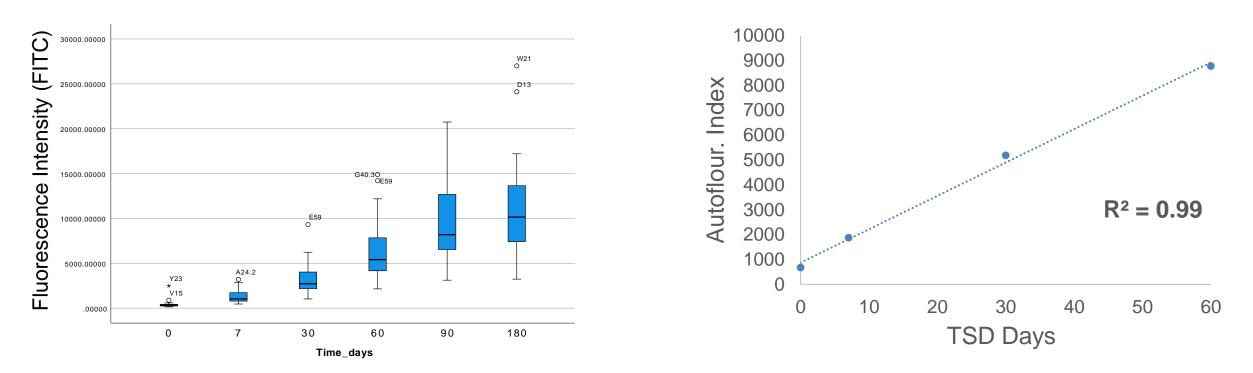
Vacant Laboratory

• Question for mixture samples: "Are there fresh cells present ?"



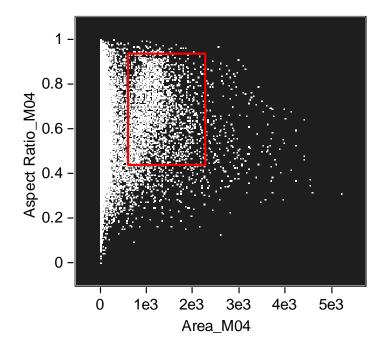
		Cell Counts		
Substrate	TSD combination	< 1 week	1 week-2 mos	> 3 months
Computer Mouse	~2 weeks / 1 day			
Gas Cap 1	> 1 year/ 1 day			
Gas Cap 2	> 1 year / 1 day			
Gas Cap 3	>1 year / 1 day			

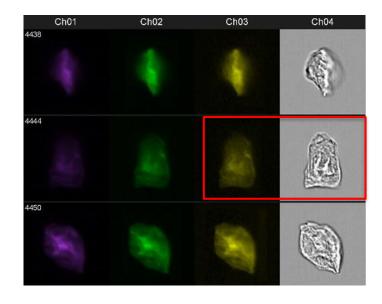
• Will this also work for other cell types, e.g. saliva?

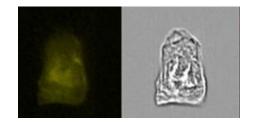


Future Directions

Research focus has been signatures themselves (not the instrument)
& quantitative framework for interpretation

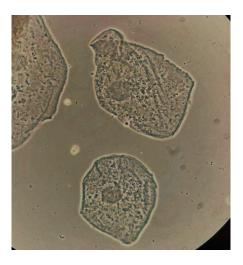




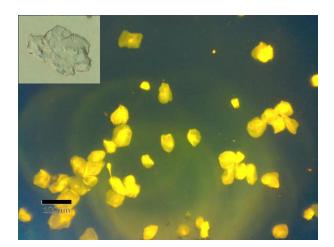


Future Directions

- TSD signatures can be obtained with <u>any</u> microscope
- Challenge is 'segmenting' cells out of the image for analysis



Zeiss Axioscope A1 (~\$10K)



Handheld fluorescent microscope (\$800)



Cell phone portable microscope (\$17)

Future Directions

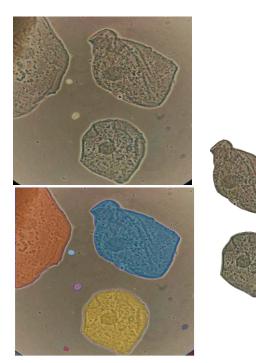
• Meta's image segmentation tool

AI Computer Vision Research

Segment Anything Model (SAM): a new AI model from Meta AI that can "cut out" any object, in any image, with a single click

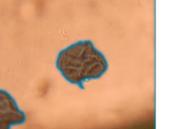
SAM is a promptable segmentation system with zero-shot generalization to unfamiliar objects and images, without the need for additional training.









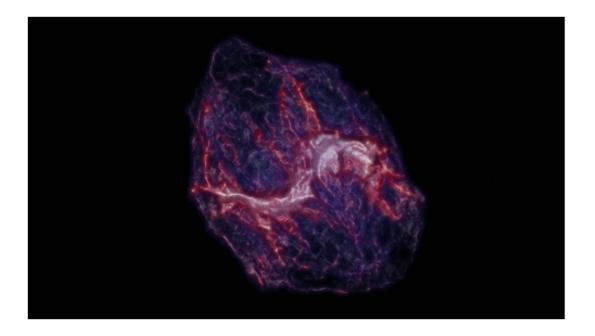


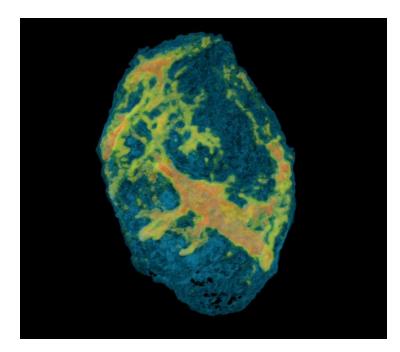


🔿 Meta

Cellular Autofluorescence

• Figuring out which compounds are causing autofluorescence and driving TSD signatures an ongoing challenge !





Laser Confocal Scanning Microscopy

Acknowledgements

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