

The current version of the human genome reference assembly, GRCh38.p14 (GRCh38), has millions of bases represented by the letter "N," which means that the actual base residing at that location is unknown.

There are also 169 sequences that cannot confidently be ordered or oriented within the assembly, typically owing to their repetitive nature

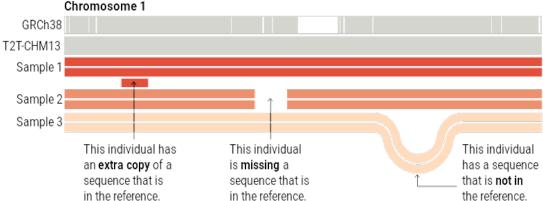
Until recently, limitations of sequencing technology, primarily that the sequencers could read no more than about 1000 bases at a time,

The HGP opted for a more structured approach. This involved cloning genomic DNA into pieces that could be grown in bacteria (clones) and indexed in 96-well plates. Clones from these libraries were first mapped to chromosome regions, then individually sequenced.

A more complete reference

The new human genome assembly, T2T-CHM13 from the Telomere-to-Telomere Consortium, includes complex and repetitive regions of chromosomes that had not been included in previous versions of the human genome assembly (GRCh38). Although the Y chromosome remains to be completed, this new reference could be annotated with regulatory regions, variants, and sequence diversity to give a fuller picture of human genomic variation.





An important attribute of the human reference assembly is that the source DNA was derived from multiple individuals.

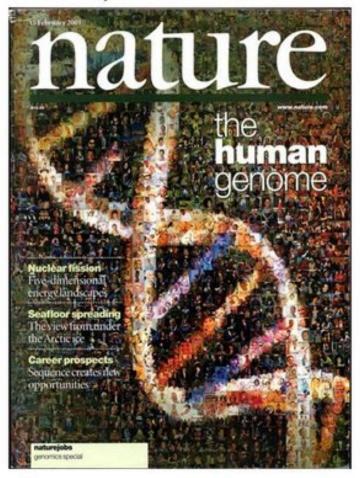
when two clones from different haplotypes of an individual are adjacent in the reference assembly, this can create sequence representations that are not normally found in the population

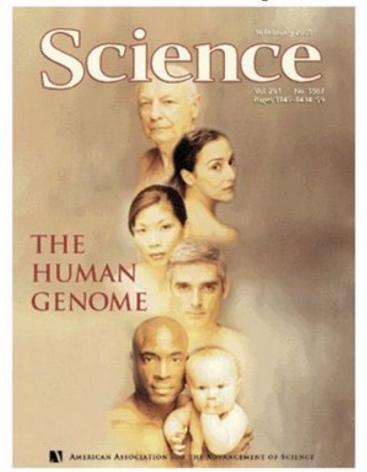
T2T Consortium, eliminated the problem of allelic diversity by sequencing the genome of a cell line derived from a complete hydatidiform mole (CHM).

This is duplicated so that the cell contains two copies of the same parental genome

Genome sequencing

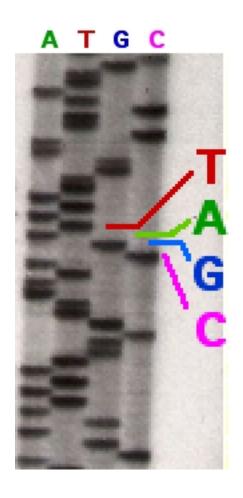
February 2001 - Publication of the first draft of the human genome

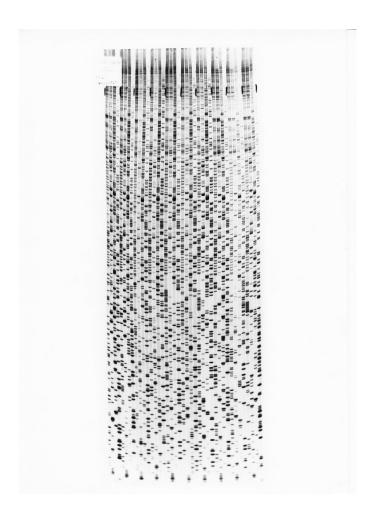




DNA Sequencing

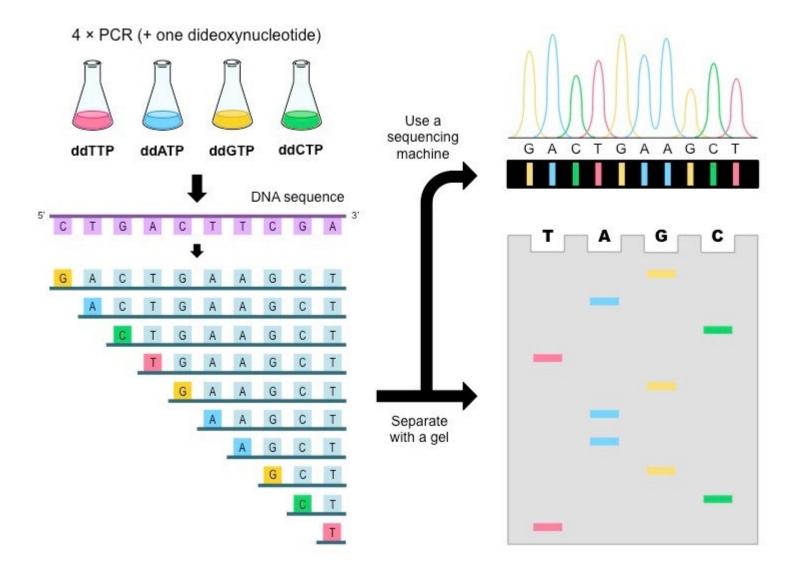
https://www.youtube.com/watch?v=vK-HIMaitnE



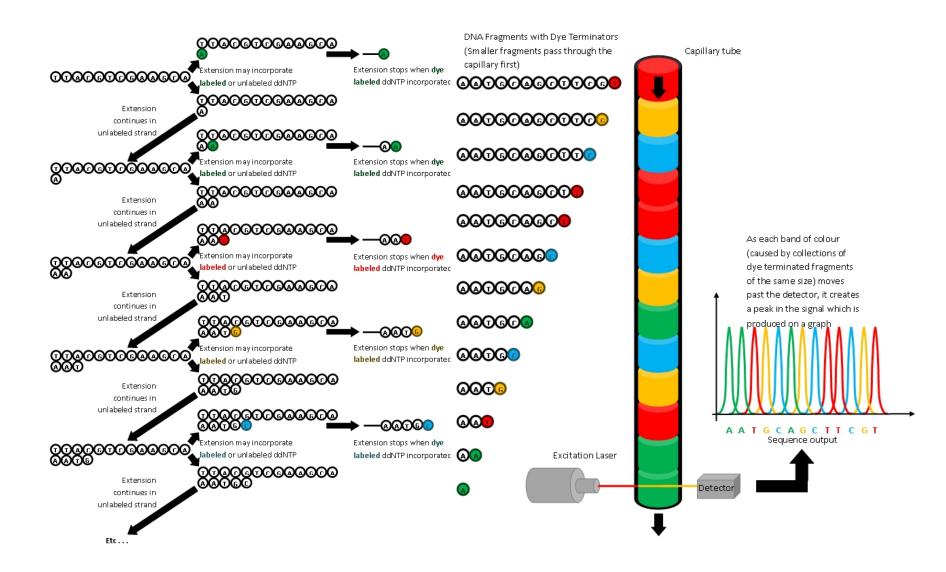


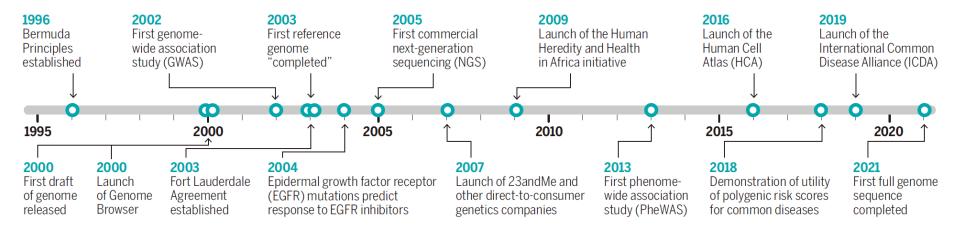
100-1000 bps

DNA Sequencing



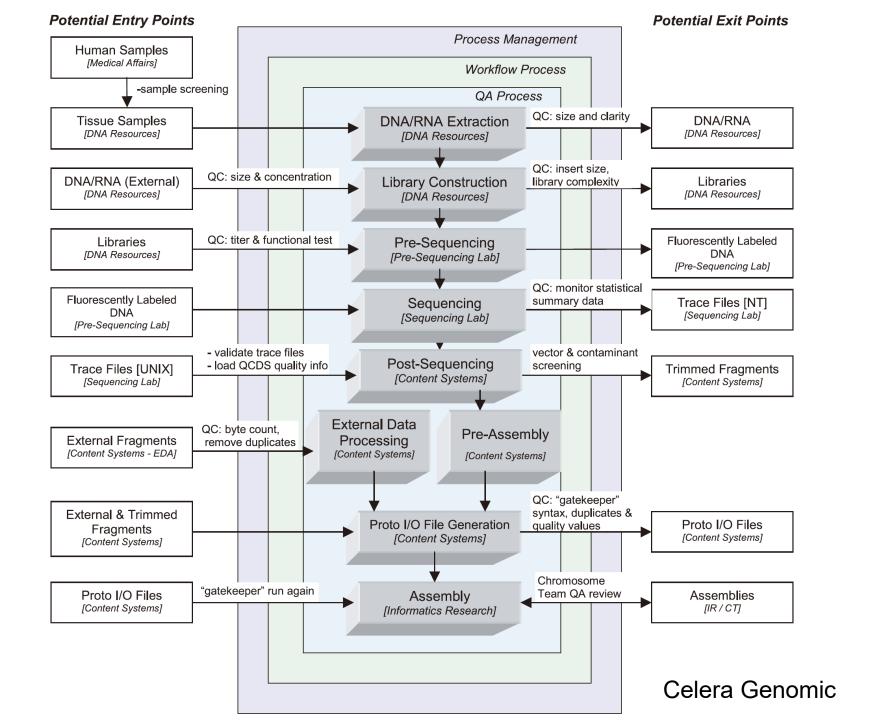
Dye Terminations

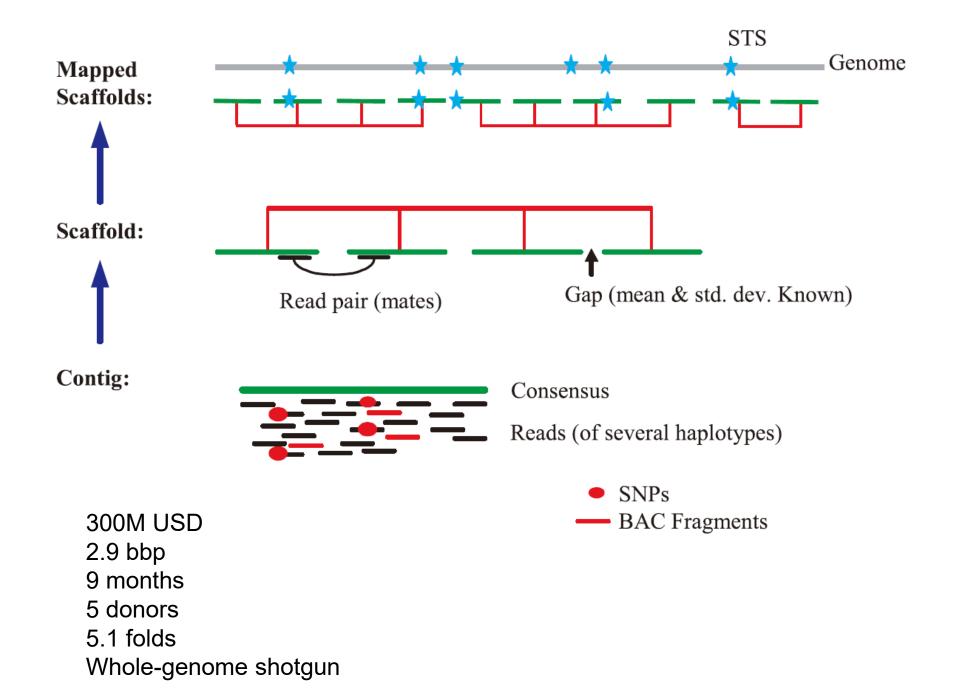




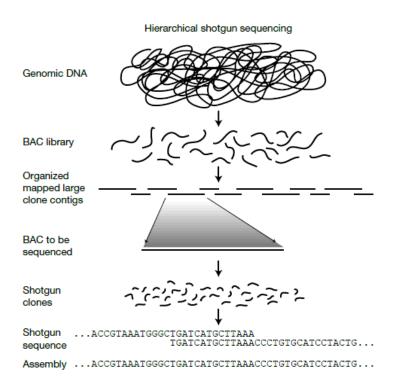
Bermuda Principles

- •Automatic release of sequence assemblies larger than 1 kb (preferably within 24 hours).
- •Immediate publication of finished annotated sequences.
- •Aim to make the entire sequence freely available in the public domain for both research and development in order to maximise benefits to society.



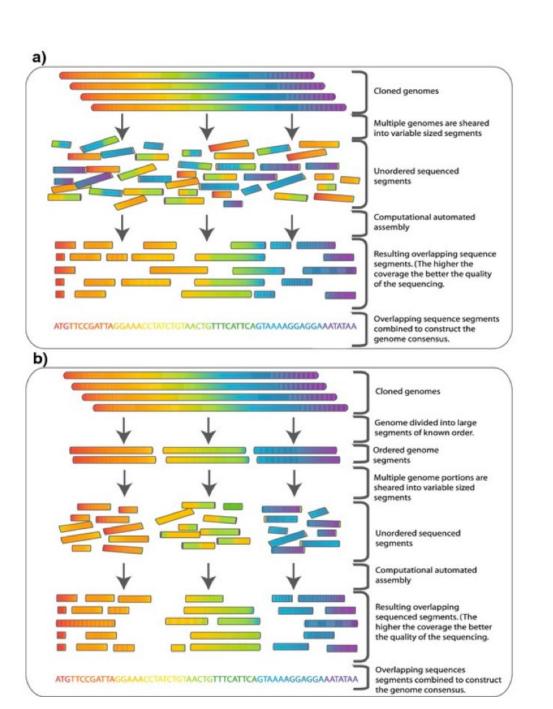


Human Genome Project



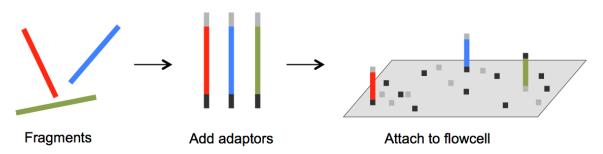


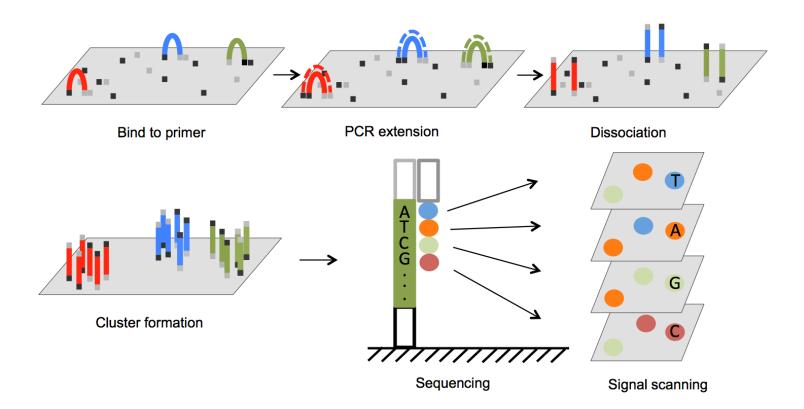
1990 15-year project 3B USD 20 groups 15 months



Next Generation Sequencing

https://www.youtube.com/watch?v=fCd6B5HRaZ8



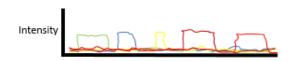


Third Generation Sequencing

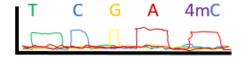
PacBio SMRT seq DNA passes thru polymerase in an illuminated volume



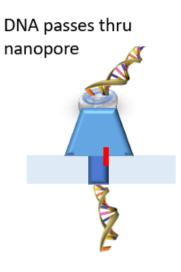
Raw output is fluorescent signal of the nucleotide incorporation, specific to each nucleotide



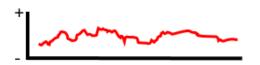
A,C,T,G have known pulse durations, which are used to infer methylated nucleotides



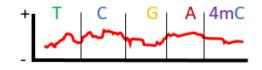
Oxford Nanopore



Raw output is electrical signal caused by nucleotide blocking ion flow in nanopore

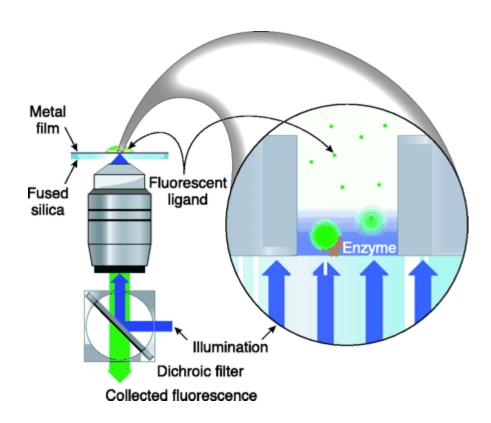


Each nucleotide has a specific electric "signature"

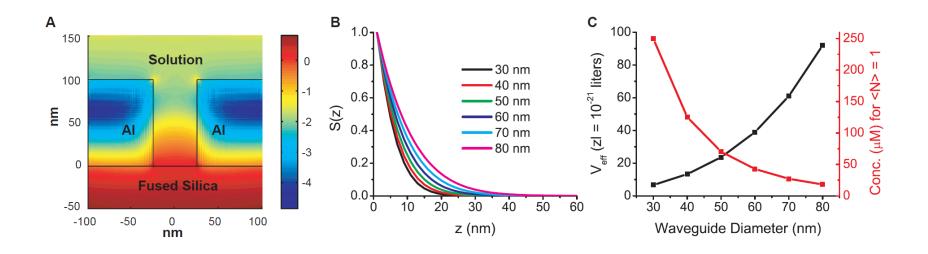


Zero Mode Waveguide

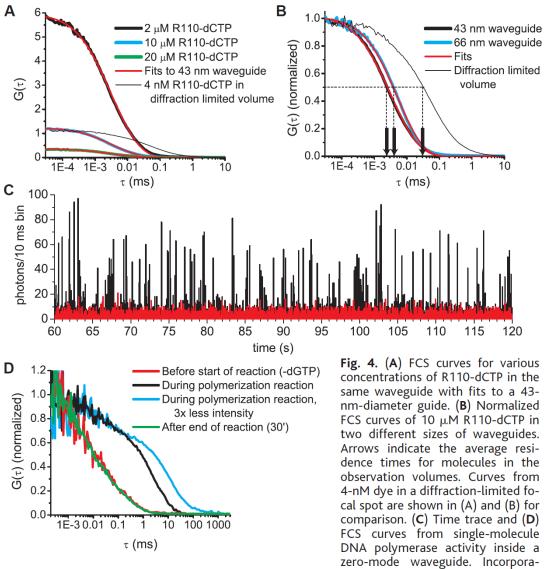
https://www.youtube.com/watch?v=NHCJ8PtYCFc



Zero Mode Waveguide

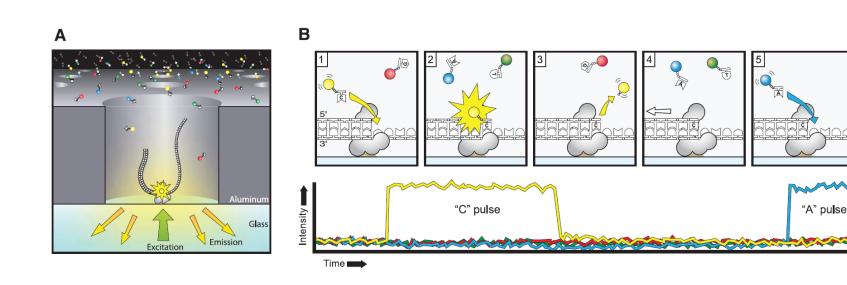


 $50 \times 50 \times 10 \text{ nm}^3 = 2.5 \times 10^4 \times 10^{-21} \text{ cc} = 2.5 \times 10^{-20} \text{ L} = 25 \text{ zeptoliter}$



tion events and subsequent photobleaching of coumarin-dCTP appear as distinct fluorescence bursts in the black time trace (10-ms time bins). This results in a long-time shoulder in the corresponding FCS curves during polymerization (black and blue curves) in (D). Decreasing the intensity results in slower photobleaching as seen by the longer residence time in the blue curve. The red curves in (C) and (D) are the corresponding negative controls (absence of one native nucleotide) in the same waveguide before initiation of the reaction. The green curve in (D) is the control after the reaction has stopped.

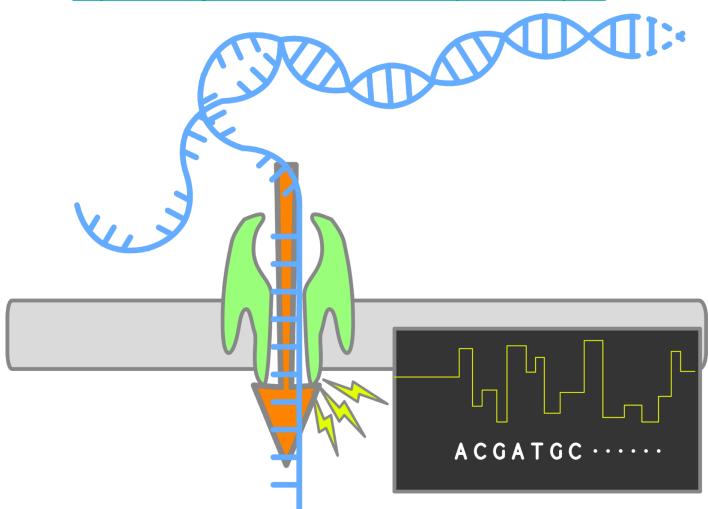
Real-Time DNA Sequencing from Single Polymerase Molecules

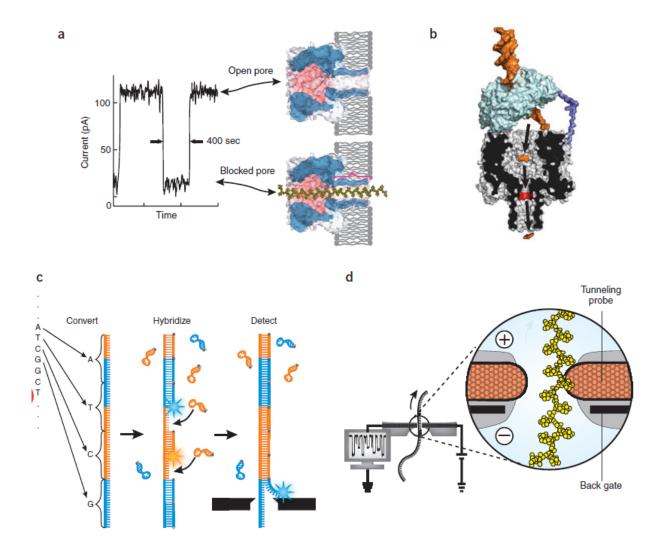


Nanopore Sequencing

https://www.youtube.com/watch?v=RcP85JHLmnI

https://www.youtube.com/watch?v=qzusVw4Dp8w





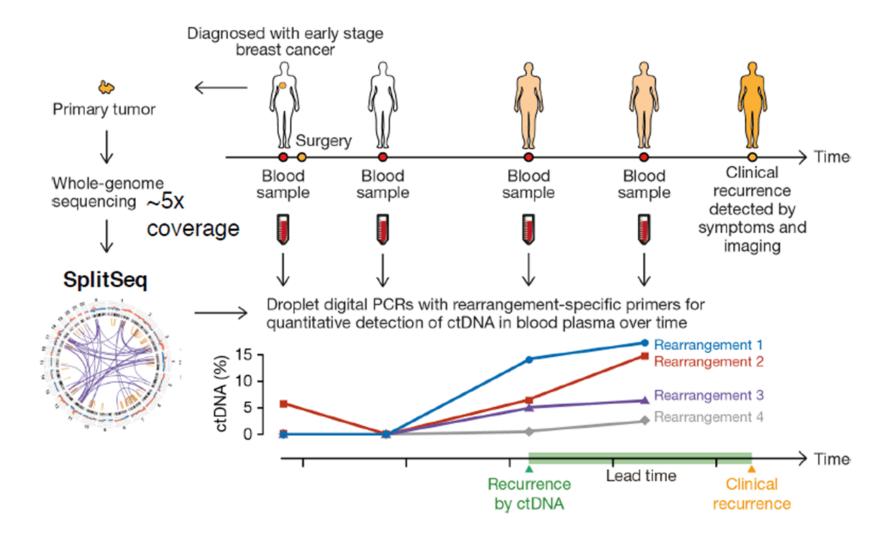
Cell Free DNA (cfDNA)

TABLE 2 Cell-free (cf)DNA concentrations and tumour response according to response evaluation criteria in solid tumours (RECIST) criteria

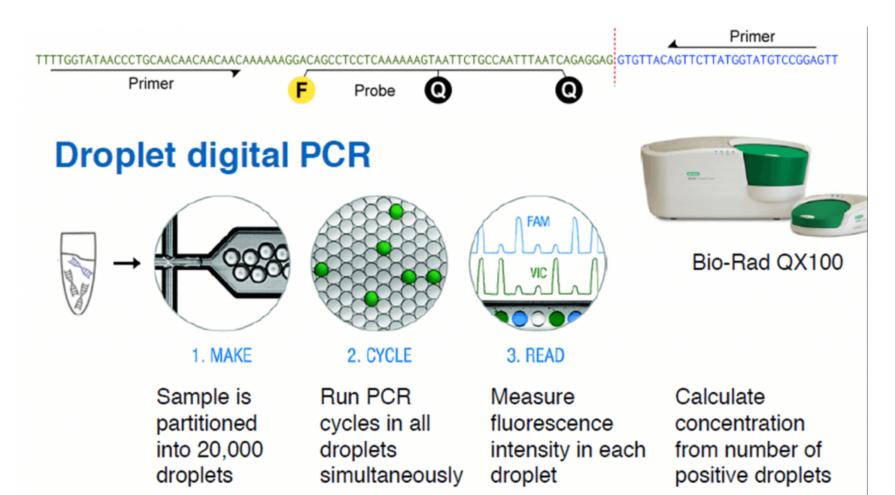
	Progressive disease	Stable disease	Partial response	p-value
Baseline concentration ng·mL ⁻¹	23.88 (35.84)	32.83 (37.32)	26.79 (28.98)	0.358
Post-chemotherapy concentration ng·mL ⁻¹	24.16 (21.66)	28.61 (37.92)	30.72 (61.33)	0.358
Difference between post-chemotherapy and baseline concentration ng·mL ⁻¹	-0.22 (27.52)	-2.01 (28.63)	-0.56 (41.95)	0.473
Variation in concentration %	-0.01 (1.04)	-0.08 (0.92)	-0.02 (1.80)	0.402

Data are presented as median (interquartile range), unless otherwise stated.

0.01%-90% circulating tumor DNA (ctDNA)



Digital PCR



Towards precision medicine

Euan A. Ashley

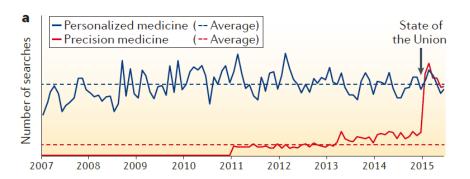
NATURE REVIEWS | GENETICS

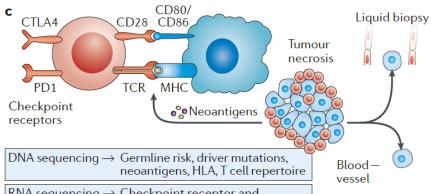
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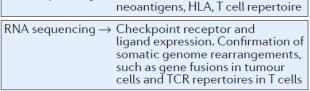
Box 1 | Personalized medicine, precision medicine and precision health

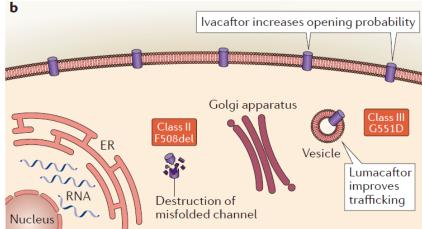
Semantically, precision and accuracy are distinct concepts. Precision reflects the extent to which repeated measurements are similar, whereas accuracy reflects the extent to which a given measurement reflects the truth. A common analogy is a target where precise but inaccurate shots cluster together away from the centre, whereas accurate but imprecise shots scatter widely around the centre. Although the US National Research Council explicitly includes the concepts of precision and accuracy in its definition of precision medicine¹⁵, and we can paint both concepts in a genomic context, neither quite captures the essence of precision medicine as currently defined. The current definition — understanding disease at a deeper level in order to develop more targeted therapy — clearly requires genomic tools that are both accurate (the genome is represented faithfully) and precise (repeating the same test multiple times leads to the same result). Notably, the US National Research Council distinguished precision medicine from personalized medicine, which it defined as the situation in which therapeutics are synthesized for specific individuals¹⁵. However, most people probably believe personalized medicine instead to mean some degree of personalization that would incorporate, for example, pharmacogenomics-based tailoring of therapy, as well as the fruits of precision medicine approaches. Finally, precision medicine is increasingly recognized as synonymous with a technology-driven and participant-centred approach. A final extension includes the concept of precision health: using similar approaches for disease prevention and health promotion.

The current definition — understanding disease at a deeper level in order to develop more targeted therapy — clearly requires genomic tools that are both accurate (the genome is represented faithfully) and precise (repeating the same test multiple times leads to the same result).

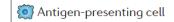








Subcategory	Functional effect	Example variant
L	No functional protein	G542x
II	Trafficking defect	F508del
III	Defective regulation	G551D
IV	Decreased conductance	R117H
V	Reduced synthesis	3120+1G>A
VI	Reduced stability	Q1412x



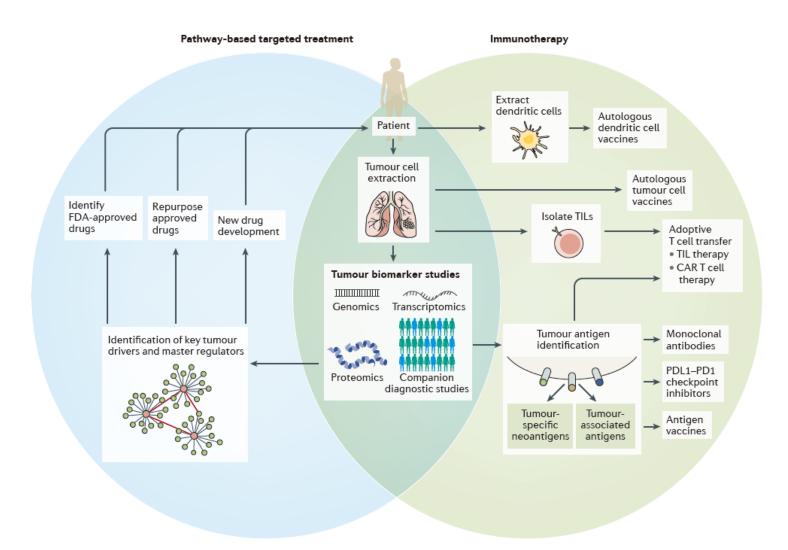




Drug development in the era of precision medicine

NATURE REVIEWS | DRUG DISCOVERY

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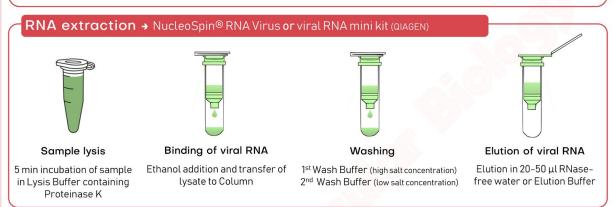


PROTOCOL OF SARS-COV-2 DETECTION USING REAL-TIME RT-PCR

Target gene → RdRp gene (Corman et al. 2020)

PCR amplification regions → nCoV_IP2/12621-12727 and nCoV_IP4/14010-14116 (Institut Pasteur. Paris)

 $\textbf{Primer sets and probes} \rightarrow \text{designed based on the first sequences of SARS-CoV-2 available on the $\textbf{GISAID database}$ }$



Real-time Multiplex RT-PCR (Institut Pasteur, Paris) Amplification Cycles (Lightcycler System) Multiplex Mix (nCoV_IP2&IP4) Denaturation amplification Cooling X 50 Sample RNA 5 μl H20 $1.3 \mu l$ 95°C 95°C 12.50 µl Reaction mix 2X MgS04 (50mM) $0.40 \mu l$ 15 sec 3 min Forward Primer1 (10µM) $1.00 \mu l$ Reverse Primer1 (10µM) $1.00 \mu l$ Reverse Forward Primer2 (10µM) $1.00 \mu l$ transcription 58°C Reverse Primer2 (10µM) 1.00 µl Probe 1 (10µM) 30 sec $0.4 \mu l$ 55°C Probe 2 (10µM) $0.4 \mu l$ 20 min SuperscriptIII RT/Platinum Taq Mix 1.00 ul 40°C 30 sec **POSITIVE CONTROL**

References

1. Institut Pasteur, Paris, « Protocol: Real-time RT-PCR assays for the detection of SARS-CoV-2 », OMS, 2 mars 2020.

The transcript contains the amplification regions of the RdRp and E gene as positive strand.

 $2. \, Corman \, VM, \, Landt \, 0. \, Kaiser \, M, \, et \, al. \, Detection \, of \, 2019 \, novel \, coronavirus \, (2019-nCoV) \, by \, real-time \, RT-PCR. \, Euro \, Surveill \, 2020; 25. \, Corman \, VM, \, Landt \, O. \, Kaiser \, M. \, et \, al. \, Detection \, of \, 2019 \, novel \, coronavirus \, (2019-nCoV) \, by \, real-time \, RT-PCR. \, Euro \, Surveill \, 2020; 25. \, Corman \, VM, \, Landt \, O. \, Kaiser \, M. \, et \, al. \, Detection \, of \, 2019 \, novel \, coronavirus \, (2019-nCoV) \, by \, real-time \, RT-PCR. \, Euro \, Surveill \, 2020; 25. \, Corman \, VM, \, Landt \, O. \, Corman \, VM, \, Cor$

Positive control for real-time RT-PCR is the in vitro transcribed RNA derived from strain BetaCoV_Wuhan_WIV04_2019.

1 hour Known primer

Hematology Analyzer



WBC White Blood Cell or leukocyte count RBC Red Blood Cell or erythrocyte count

Hgb Hemoglobin concentration

Hct Hematocrit (relative volume of erythrocytes)
MCV Mean Corpuscular (erythrocyte) Volume
MCH Mean Corpuscular (erythrocyte) Hemoglobin

MCHC Mean Corpuscular (erythrocyte) Hemoglobin Concentration

RDW Red Cell (erythrocyte volume) Distribution Width

RDW-SD Red Cell (erythrocyte volume) Distribution Width (Standard Deviation)

Plt Platelet or thrombocyte count

MPV Mean Platelet (thrombocyte) Volume

LY% Lymphocyte percent **MO%** Monocyte percent NE% Neutrophil percent E0% Eosinophil percent BA% Basophil percent LY# Lymphocyte number MO# Monocyte number NE# Neutrophil number EO# Eosinophil number Basophil number BA#

NRBC% Nucleated Red Blood Cell percent
NRBC# Nucleated Red Blood Cell number

RET% Reticulocyte percent
RET# Reticulocyte number

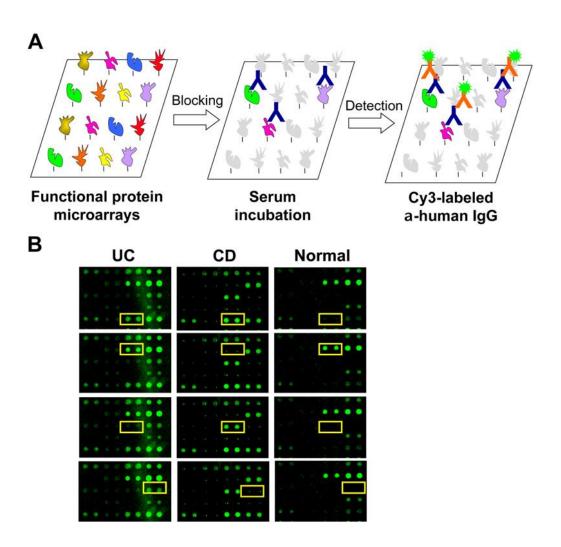
IRF Immature Reticulocyte Fraction
MRV Mean Reticulocyte Volume

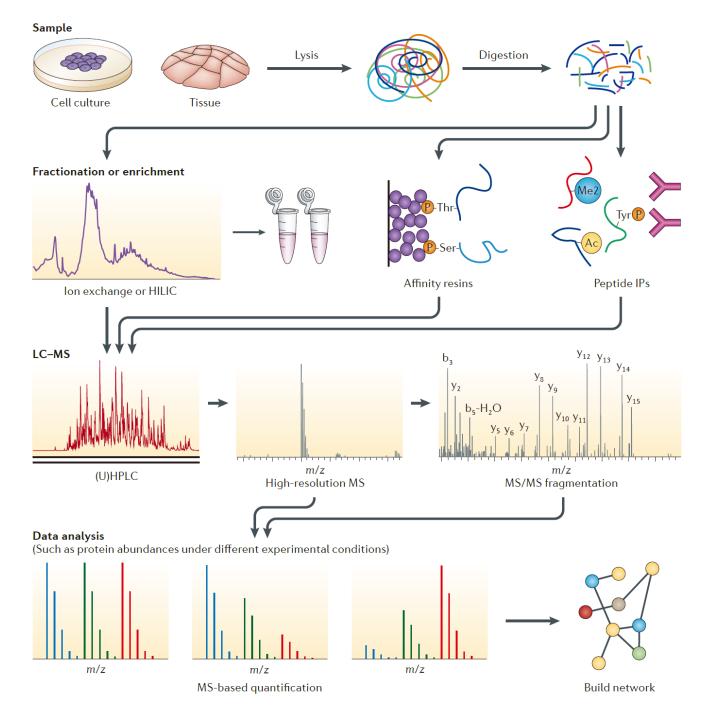
Chemistry Analyzer

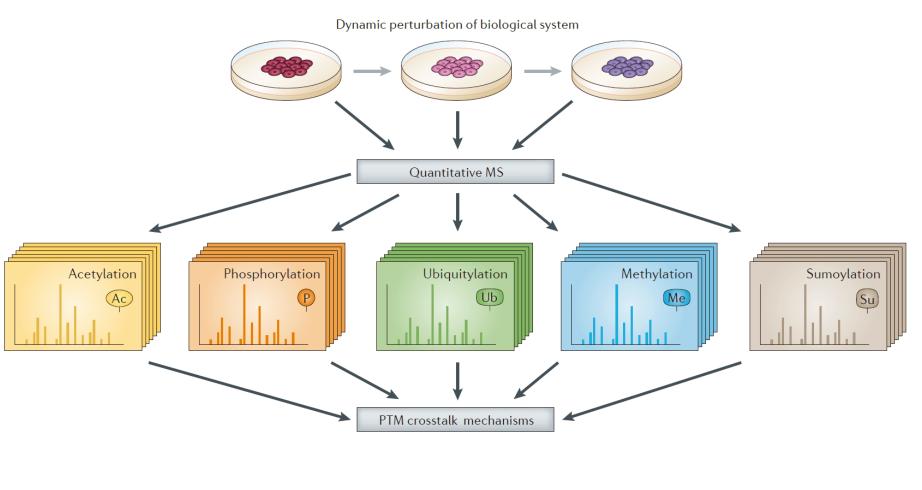


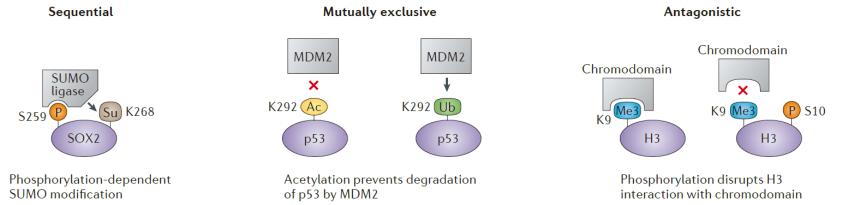
No.	Test Name	Long Name
1	ALB	Albumin
2	ALP	Alkaline Phosphatase
3	ALT	Alanine Transferase
4	AMY	Amylase
5	AST	AspartateTransferase
6	C02	Bicarbonate
7	DBILC	Direct Bilirubin
8	DBB	DirectBilirubinBlank
9	TBILC	Total Bilirubin
10	TBB	TotalBilirubinBlank
11	CA	Calcium Arsenazo
12	NDBILI	Direct Bili - 181
13	CHOL	Cholesterol
14	AMMON	Ammonia
15		
16	CK	CreatPhosphoKinase
17	CRE	Creatinine
18	GGT	Gamma Glutamyl Trans
19	GLU	Glucose

Protein Microarray









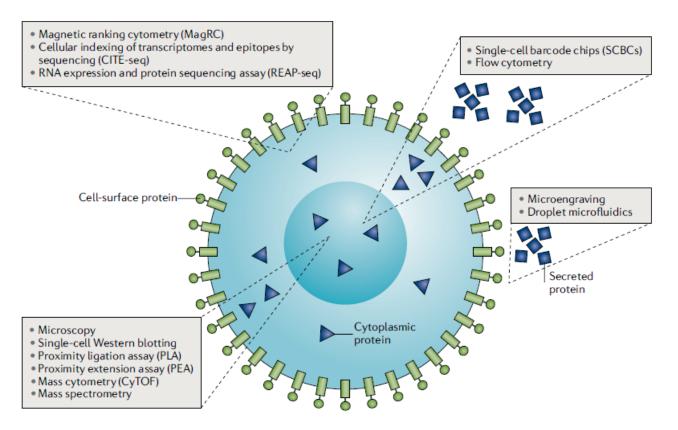
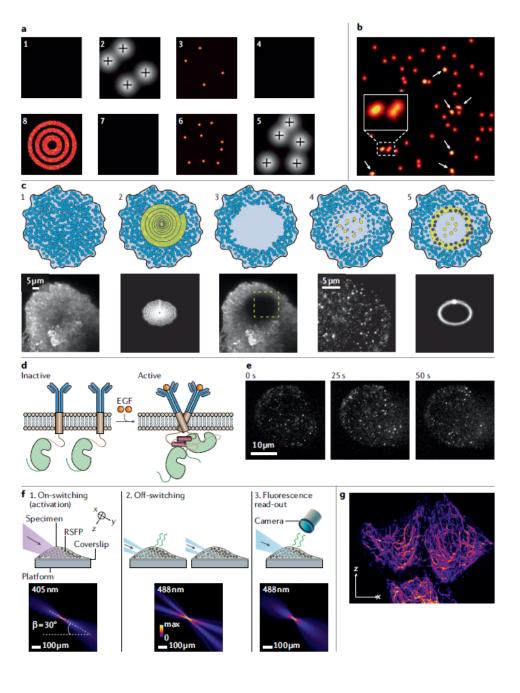
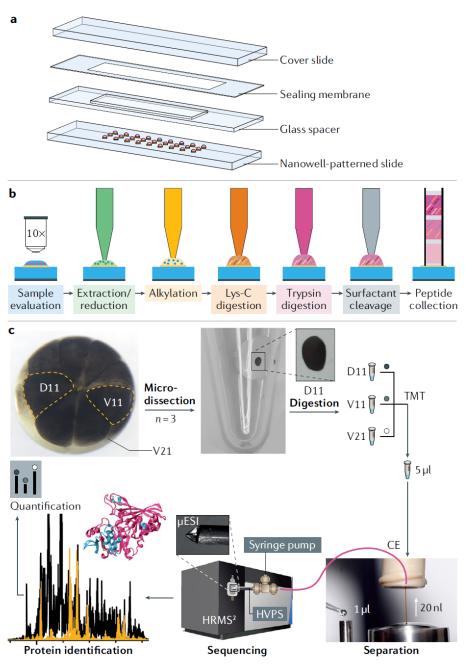
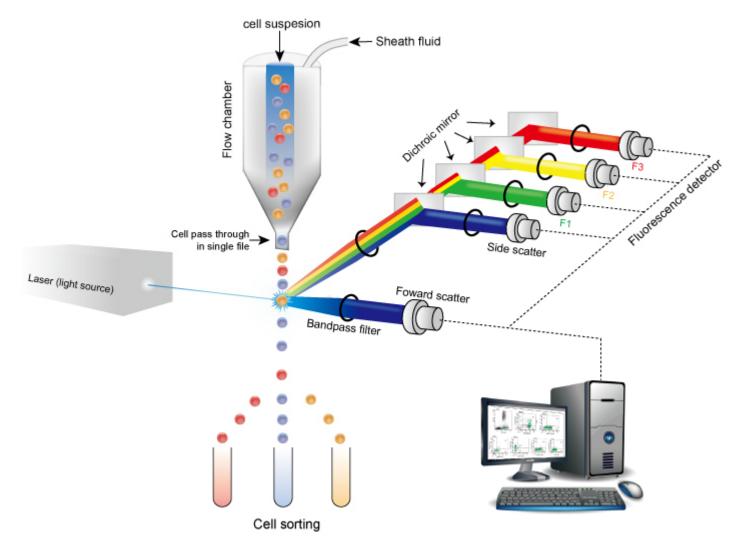


Fig. 1 | Classification of single-cell protein analysis methods based on the location of target protein. Cell-surface-protein analysis methods include magnetic ranking cytometry (MagRC), cellular indexing of transcriptomes and epitopes by sequencing (CITE-seq), and RNA expression and protein sequencing assay (REAP-seq). Methods that can be used for the analysis of cell-surface and cytoplasmic proteins include microscopy, single-cell Western blotting, proximity ligation assay (PLA), proximity extension assay (PEA), mass cytometry (cytometry by time of flight; CyTOF) and mass spectrometry. Methods utilized for secreted-protein analysis include droplet microfluidics and microengraving techniques. Methods used for comprehensive analysis of the three proteins include flow cytometry and single-cell barcode chips (SCBCs).

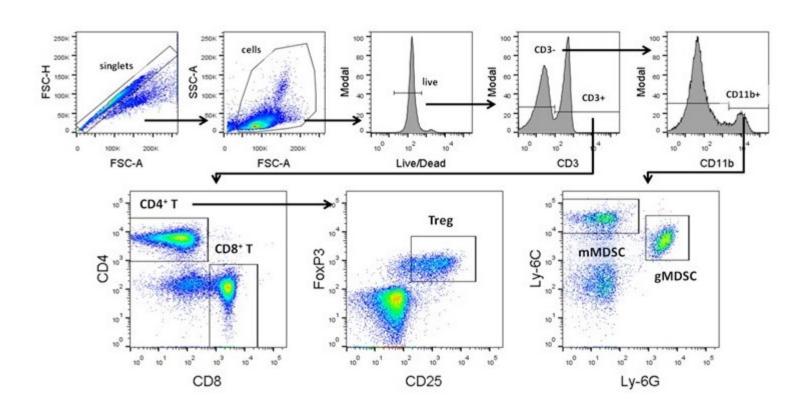




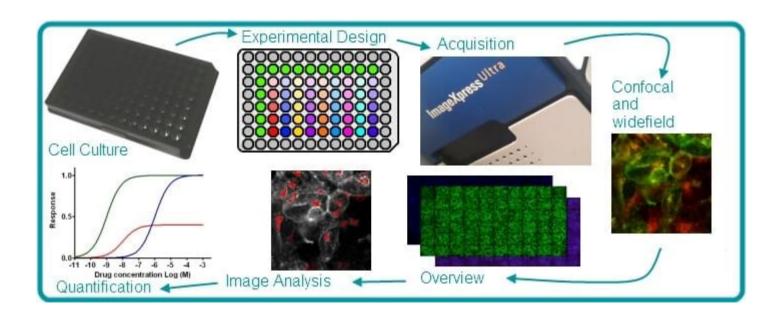
Flow Cytometry



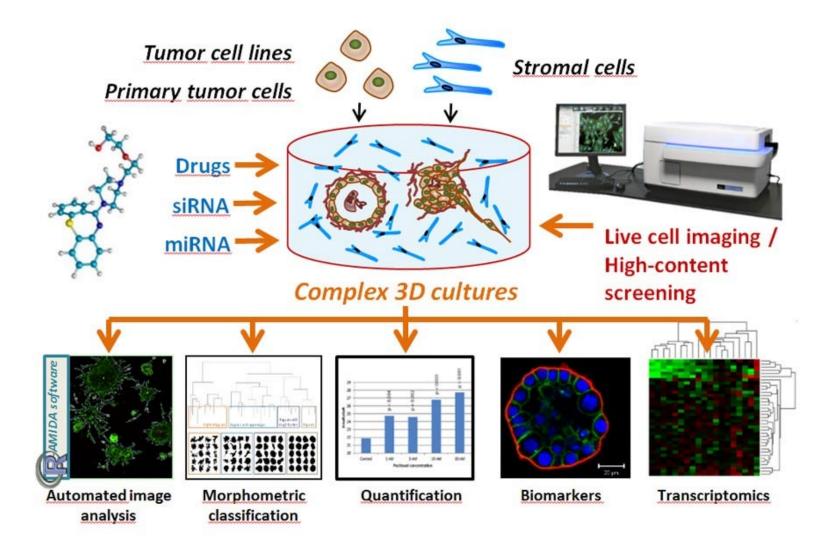
Flow Cytometry Data



High Content Screening



High Content Screening



Single Cell Analysis

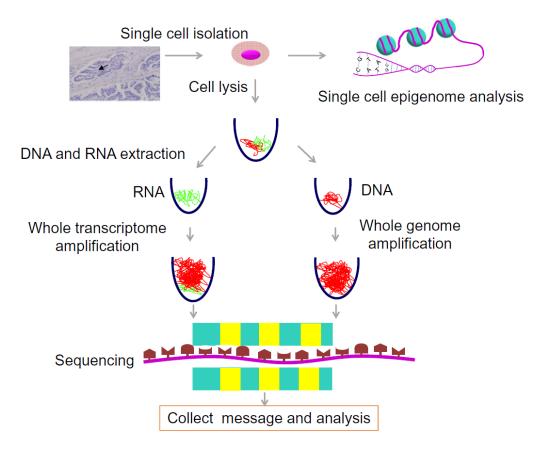
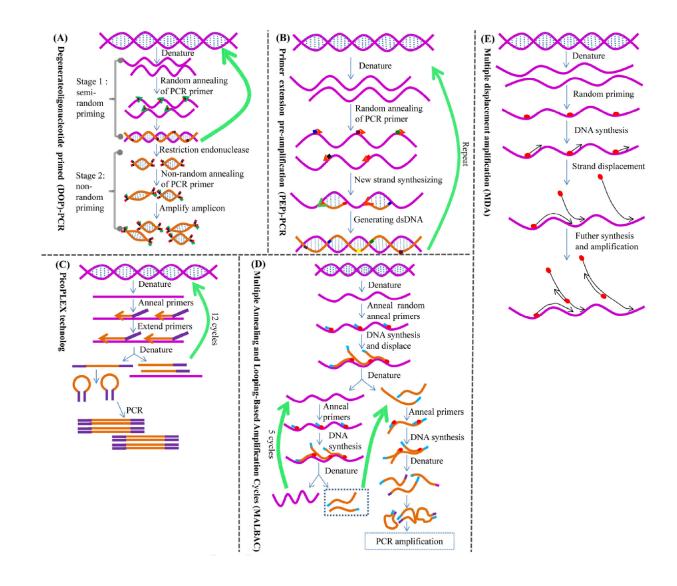
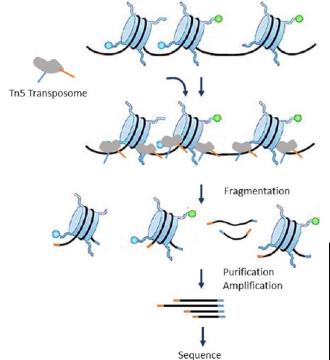


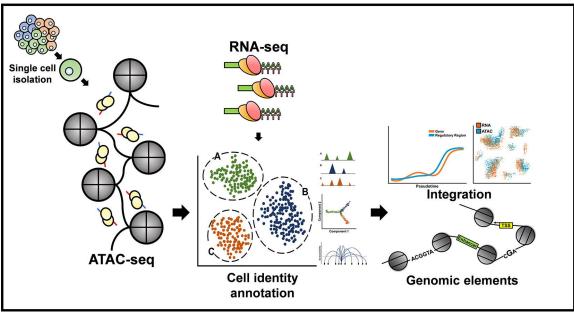
Fig. 1. Single-cell sequencing of a tumor cell. A tumor specimen is obtained by surgical excision and single cells are isolated by one of the several methods shown in Fig. 2. The individual cancer cell can be used for epigenome sequencing directly or lysed to extract the genetic material (DNA and RNA), which is in turn amplified by the methods shown in Fig. 3. Then, the amplified DNA and RNA are sequenced by single-cell sequencing technology and the result data are analyzed to provide insights into the molecular mechanisms underlying intratumor heterogeneity.

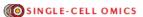
Whole Genome Amplifications



ATAC-seq (Assay for Transposase-Accessible Chromatin using sequencing)



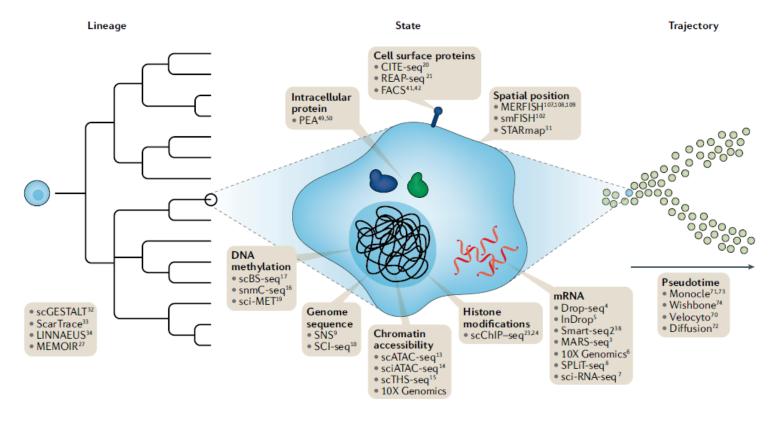




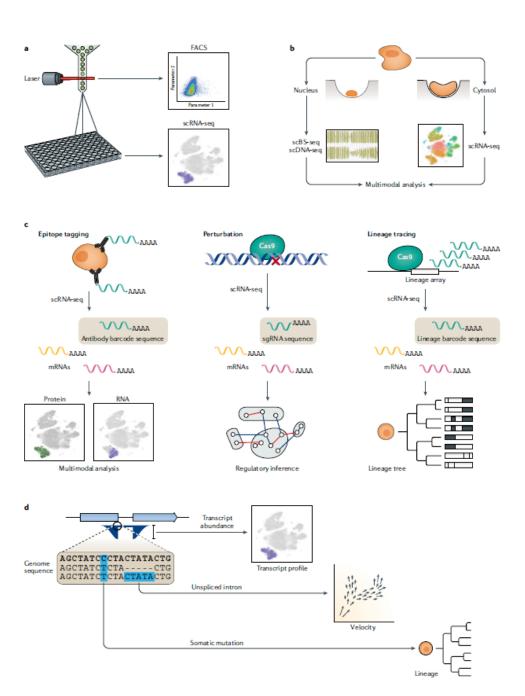
Integrative single-cell analysis

Tim Stuart 101 and Rahul Satija 101,2*

Abstract | The recent maturation of single-cell RNA sequencing (scRNA-seq) technologies has coincided with transformative new methods to profile genetic, epigenetic, spatial, proteomic and lineage information in individual cells. This provides unique opportunities, alongside computational challenges, for integrative methods that can jointly learn across multiple types of data. Integrated analysis can discover relationships across cellular modalities, learn a holistic representation of the cell state, and enable the pooling of data sets produced across individuals and technologies. In this Review, we discuss the recent advances in the collection and integration of different data types at single-cell resolution with a focus on the integration of gene expression data with other types of single-cell measurement.

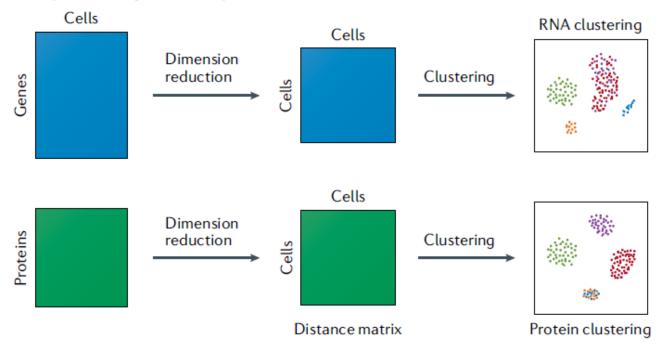


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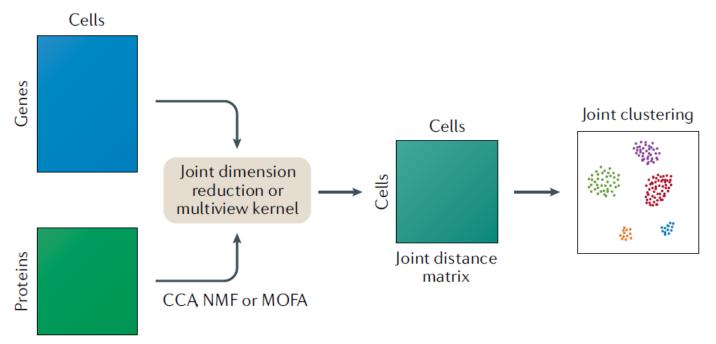
Computational Analysis for Multiomics

a Separate analysis of multiple modalities



Computational Analysis for Multiomics

b Joint analysis of multiple modalities



c Multimodal modelling framework

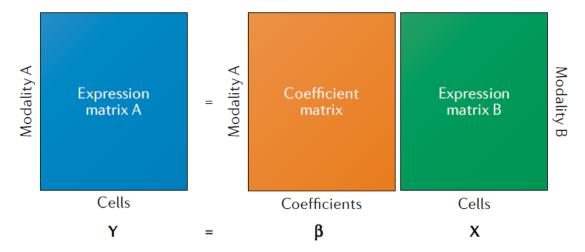
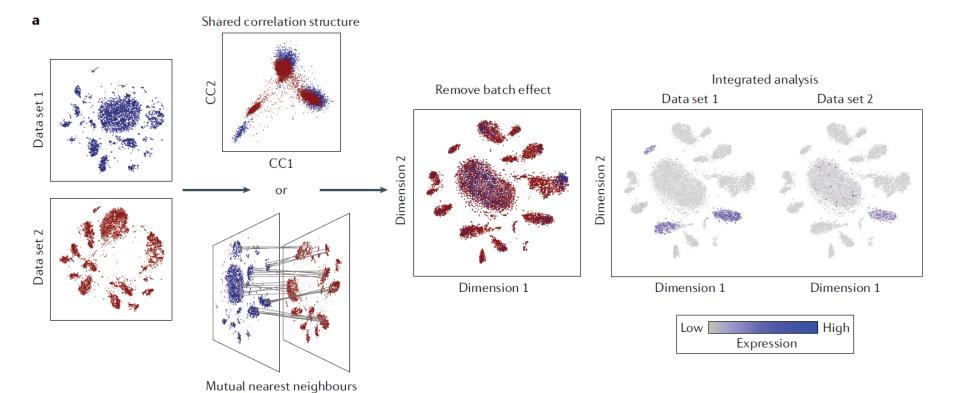
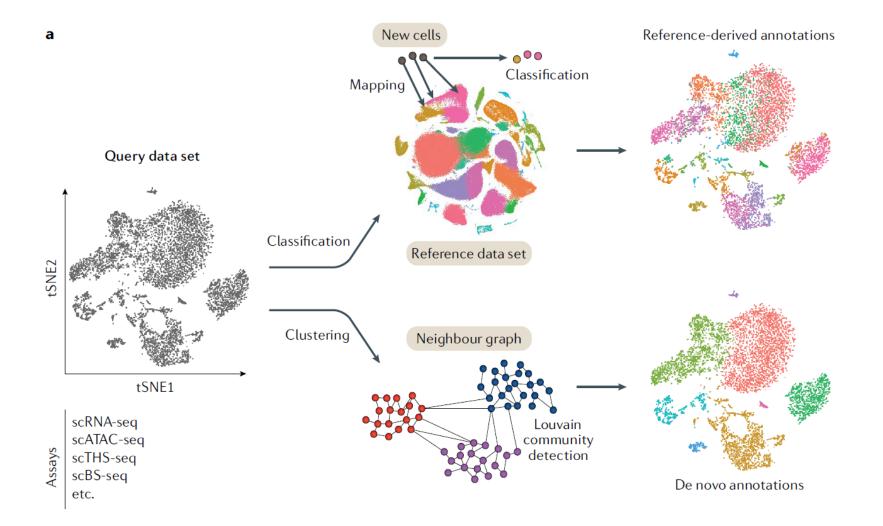
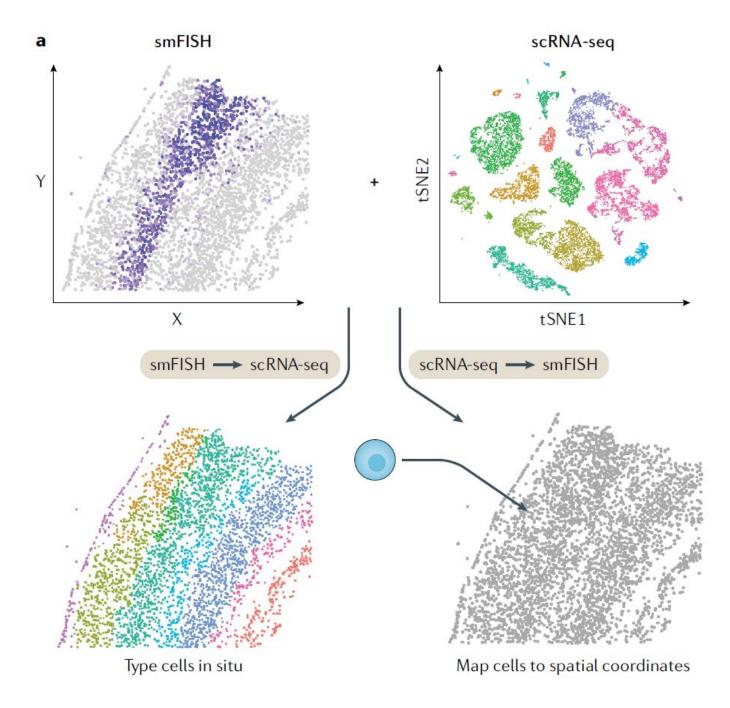


Fig. 3 | Computational methods for the analysis of multimodal single-cell data. a | Independent analysis of multiple modalities measured from the same group of cells can lead to conflicting identification of clusters of cells. Measurements from one modality may indicate that certain cells are highly similar, while the same cells appear very different when looking at a different modality. In this example, the orange and blue cells form separate clusters when analysed on the basis of their transcriptional profile but cluster together when analysed according to protein measurements. Similarly, the red and purple cells group together based on RNA but separate based on protein. **b** Joint analysis of multiple modalities measured from the same group of cells can have greater power to identify unique cell states. By taking into account information from multiple modalities, cells can be jointly clustered in a way that outperforms the classifications derived from any one modality alone. In this example, both the red and purple cells and the orange and blue cells form separate groups when protein and RNA information is analysed jointly. **c** | The relationships between data modalities can be studied by building statistical models that aim to explain variance in one modality using a linear combination of components of another modality. CCA, canonical correlation analysis; MOFA, multi-omics factor analysis; NMF, non-negative matrix factorization.

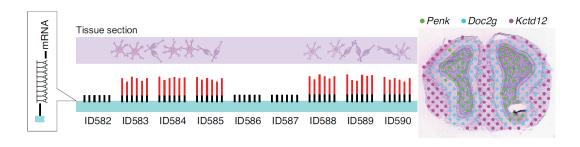




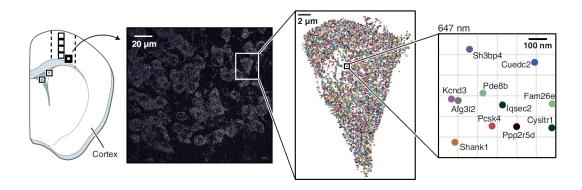


Method of the Year 2020: spatially resolved transcriptomics

Spatially resolved transcriptomics methods are changing the way we understand complex tissues.



Researchers in Sweden developed an approach in which fixed, stained tissue is imaged, permeabilized and the mRNAs attach to an array of barcoded oligos. The RNAs are reverse-transcribed; the cDNAs are sequenced and yield spatially resolved transcriptomic information. Credit: Adapted with permission from ref. ⁴, AAAS

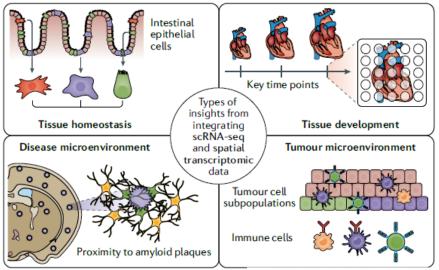


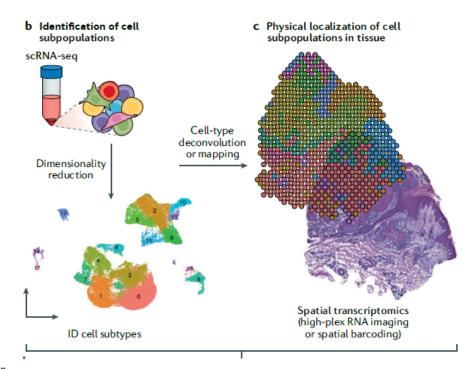
Spatial techniques help with atlas-building by localizing expressed genes. Here, seqFISH+ was used to measure 10,000 genes in mouse cortex. Credit: Cai lab, Caltech, I. Strazhnik; adapted with permission from ref. ⁶, Springer Nature.

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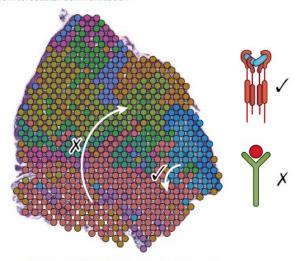
Integrating single-cell and spatial transcriptomics to elucidate intercellular tissue dynamics

a Spatial transcriptomic experimental focuses





d Decode intercellular communication



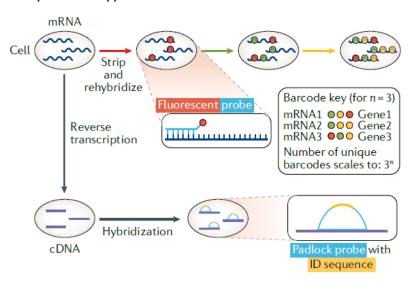
Spatially informed ligand-receptor algorithms

 $\label{thm:continuous} \mbox{Table 1} \ | \ \mbox{Widely used spatial transcriptomics techniques and instruments}$

Туре	Technique	Characteristics	Mechanism notes
High-plex RNA imaging	In situ sequencing for RNA analysis in preserved tissue and cells	25–90 reads per cell ¹⁰ , 69-gene panel ³¹	Genes of interest must be selected to design gene target-specific 'padlock probes'
	STARmap	160-gene panel, ~250 reads per cell	Uses 'padlock probe' mechanism and integrates hydrogel-tissue chemistry for 3D spatial resolution
	Multiplexed error-robust FISH (MERFISH)	135-gene panel, ~100 reads per cell ²⁹	Each gene has an associated binary code (wherein 1 means fluorescence for a certain part of the sequence, and 0 means no fluorescence)
	Sequential FISH (seqFISH, seqFISH+)	249-gene panel, ~30 reads per cell ¹³	Each gene has an associated colour sequence code (24 colour probes per gene) with 60 different pseudo-colour options
Spatial barcoding	Spatial transcriptomics	100 μm diameter capture spot resolution	100 μm from original spatial transcriptomics publication ¹⁵
	Visium spatial gene expression	55 μm diameter	55 μm resolution translates to 3–30 cells per capture spot ¹⁶ (original technique ¹⁵ is widely accessible through this 10× Genomics Visium Platform: spatialtran- scriptomics.com)
	High-definition spatial transcriptomics (HDST)	2 μm diameter	Direct improvement in resolution on spatial transcriptomics ¹⁵ , but not as accessible
	Slide-seq, Slide-seq v2	10 μm diameter	Uses beads instead of capture spots, with the poly-T oligomers projecting radially around the bead; employs slides

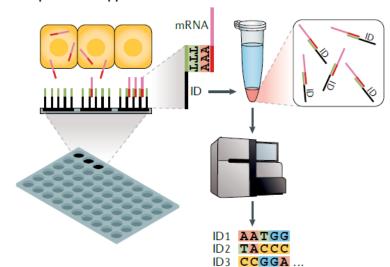
A High-plex RNA imaging

a Experimental approach

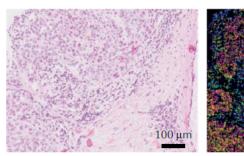


B Spatial barcoding

a Experimental approach



b Localized transcripts coloured by specific gene



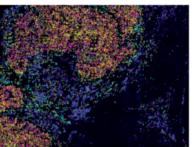
Strengths

Greater depth (per transcript)

Better suited to capture subtype

change due to spatial influence

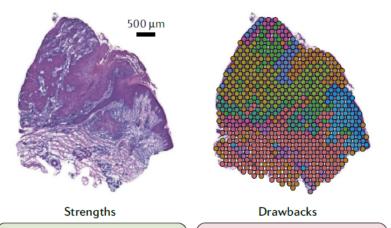
• Single-cell resolution



Drawbacks

- Biased (pre-selected gene targets required)
- Lower coverage
- Smaller field of view
- More read-out noise
- Requires more specialized equipment

b Capture spot transcript mixtures deconvolved by dominant cell type



- Unbiased
- Greater coverage
- Greater field of view
- More accessible (typically sequenced using standard NGS machine)
- Limited to capture spot resolution
- Lower depth (per transcript)

Single-cell atlases: shared and tissue-specific cell types across human organs a Tissue procurement c Enrich

a Tissue procurement **c** Enrichment d Sample multiplexing Density gradient centrifugation Plasma-Mononuclearcells RBCs-Hashtag 3 Magnetic cell sorting Sample purity Biobanks Organ donations Biopsies/resections Fluorescence- or magnetic-based **b** Tissue dissociation cell sorting Detector Temperature Time Enzyme













Neural cells (from CNS, PNS vs ENS)

Myocytes (from skeletal versus cardiac muscle)

Epithelial cells (e.g. from lung, gut)

Fibroblasts

Immune cells

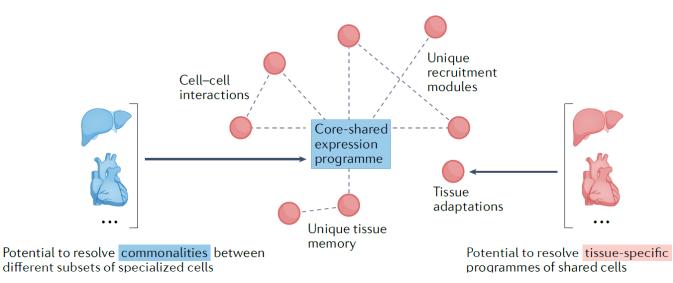
Endothelial and vascular cells

Primary cells

Generally perform organ-specialized functions

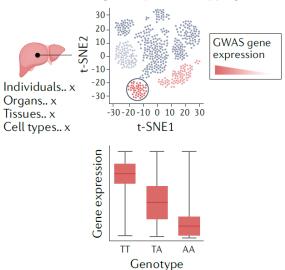
Supportive cells

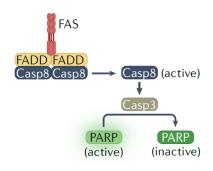
Generally perform supportive functions and are omnipresent across all organs



Ba Tissue-specific disease mechanisms

GWAS gene expression mapping

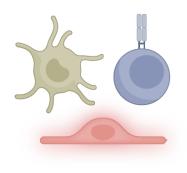




Cell-type specific and diseaseactivated pathways



Druggable cell-cell interactions

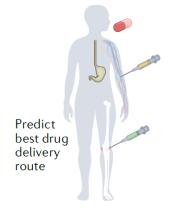


Responsive cell types based on drug target expression

Bb Therapeutic avenues

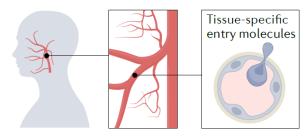


Stratify patients and responses





Diagnosis and disease monitoring based on cell type markers



Target tissue-resident immune cells

Super-Resolution Microscopy



The Nobel Prize in Chemistry 2014



Photo: Matt Staley/HHMI Eric Betzig Prize share: 1/3



Planck-Institut
Stefan W. Hell
Prize share: 1/3

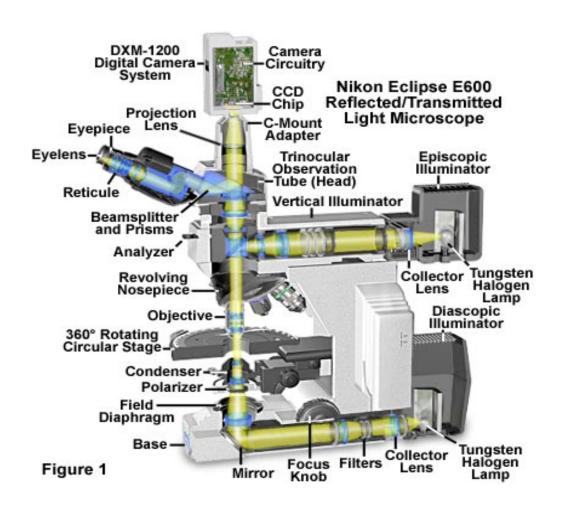


Photo: K. Lowder via Wikimedia Commons, CC-BY-SA-3.0 William E. Moerner Prize share: 1/3

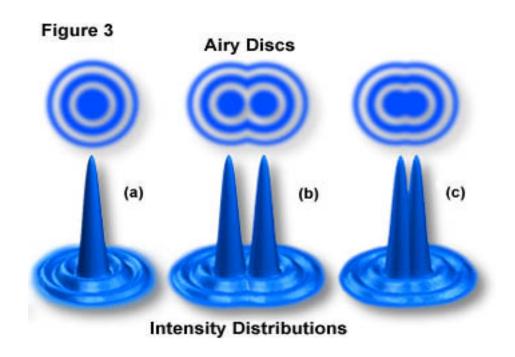
The Nobel Prize in Chemistry 2014 was awarded jointly to Eric Betzig, Stefan W. Hell and William E. Moerner "for the development of super-resolved fluorescence microscopy".

Two separate principles are rewarded. One Eric Betzig and William Moerner, working enables the method *stimulated emission* separately, laid the foundation for the *depletion (STED) microscopy,* developed bysecond method, *single-molecule microscopy.* Stefan Hell in 2000.

Microscope

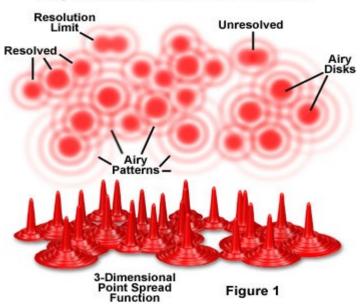


Resolution



Resolution (r) =
$$\chi$$
/(2NA) (1)
Resolution (r) = 0.61χ /NA (2)
Resolution (r) = 1.22χ /(NA(obj) + NA(cond)) (3)

Airy Patterns and the Limit of Resolution



	Objective Type					
	Plan Achromat		Plan Fluorite		Plan Apochromat	
Magnification	N.A.	Resolution (µm)	N.A.	Resolution (µm)	N.A.	Resolution (µm)
4x	0.10	2.75	0.13	2.12	0.20	1.375
10x	0.25	1.10	0.30	0.92	0.45	0.61
20x	0.40	0.69	0.50	0.55	0.75	0.37
40x	0.65	0.42	0.75	0.37	0.95	0.29
60x	0.75	0.37	0.85	0.32	0.95	0.29
100x	1.25	0.22	1.30	0.21	1.40	0.20
N.A. = Numerical Aperture						

Resolution versus Wavelength

Wavelength (Nanometers)	Resolution (Micrometers)
360	.19
400	.21
450	.24
500	.26
550	.29
600	.32
650	.34
700	.37

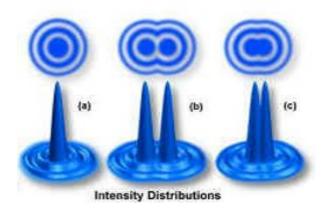
N.A. = 0.95

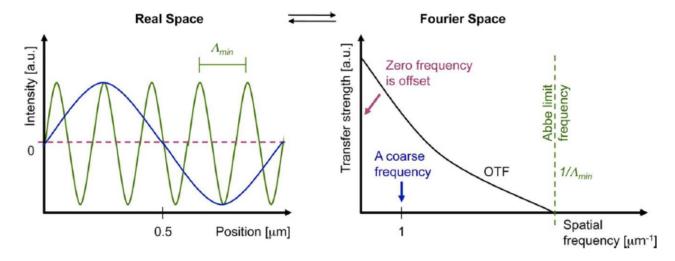
Diffraction Limit



$$d = \lambda/(2n\sin\alpha)$$

$$k_0 = 2NA/\lambda_{\rm em}$$





For a homogeneous medium, the solution to the homogeneous equation is of the form

$$H_{\omega}(k_r, z) = A^+(k_r) e^{ik_z z} + A^-(k_r) e^{-ik_z z}$$
 (2.98)

with k_z being the vertical wavenumber,

$$k_z = \sqrt{k^2 - k_r^2}. (2.99)$$

Since the inverse Hankel transform must be evaluated over a semi-infinite wavenumber domain, we have to choose a definition for the square root for $k_r > k$. We choose the definition

$$k_z = \begin{cases} \sqrt{k^2 - k_r^2}, & k_r \le k \\ i\sqrt{k_r^2 - k^2}, & k_r > k. \end{cases}$$
 (2.100)

Breaking the Diffraction Barrier: Super-Resolution Imaging of Cells

Bo Huang,1 Hazen Babcock,2 and Xiaowei Zhuang2,3,*

Cell 143, December 23, 2010 @2010 Elsevier Inc. 1047

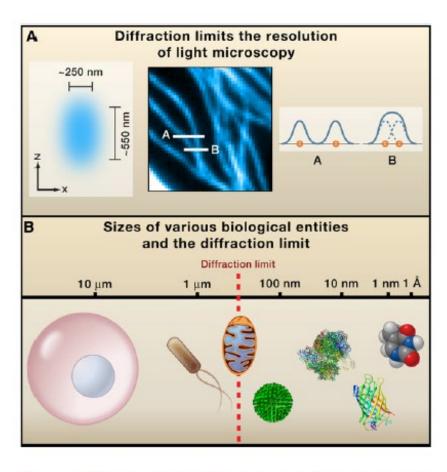


Figure 1. Diffraction-Limited Resolution of Conventional Light Microscopy

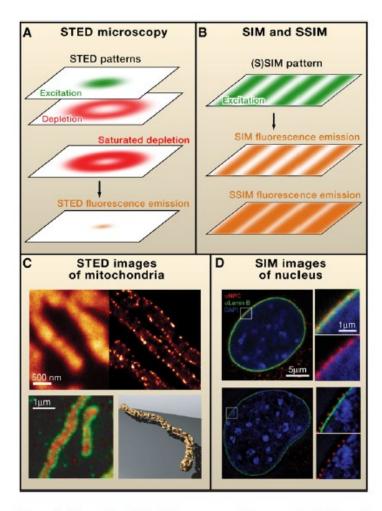


Figure 2. Super-Resolution Fluorescence Microscopy by Patterned Illumination

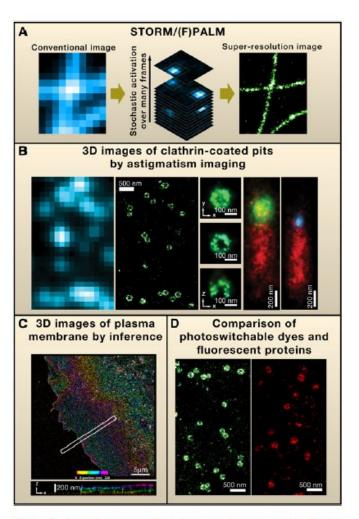
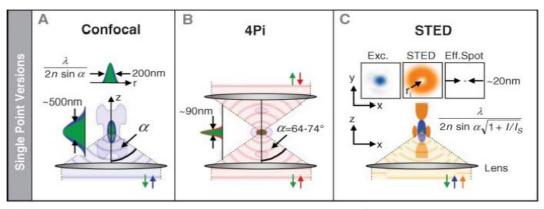
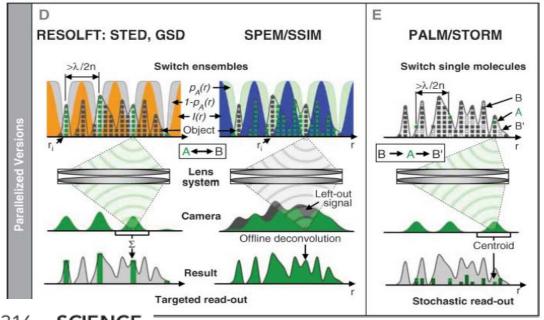


Figure 3. Super-Resolution Fluorescence Microscopy by Single-Molecule Switching

Super-Resolution Microscopy





25 MAY 2007 VOL 316 SCIENCE

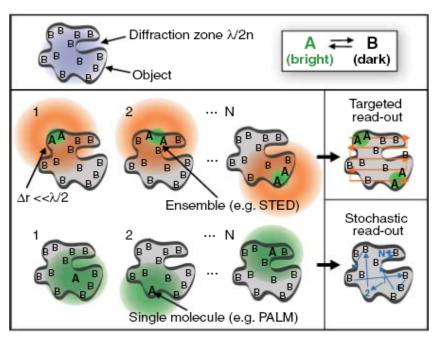
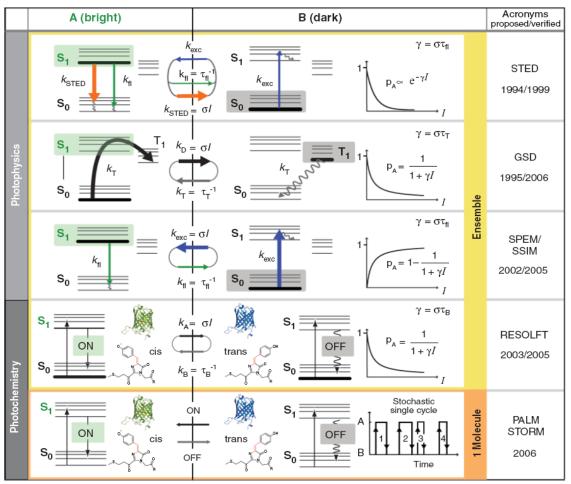
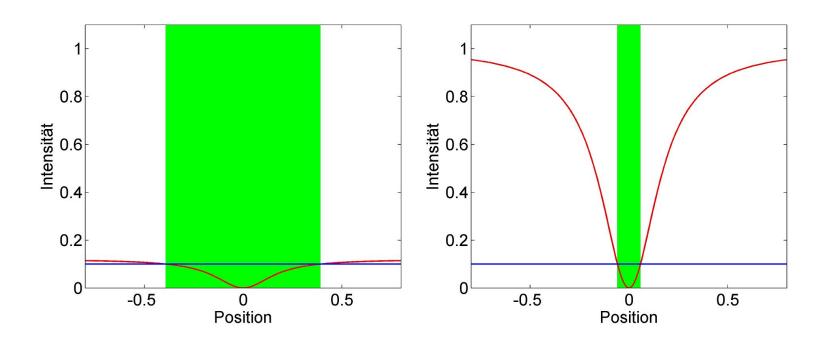


Fig. 2. Targeted versus stochastic time-sequential readout of fluorophore markers of a nanostructured object within the diffraction zone whose lower bound is given by $\lambda/2n$. A and B denote a bright and a dark state, respectively. In the targeted readout mode, one of the two states (here A) is established at a subdiffraction-sized spot at the position of a zero to read out an unknown number of fluorophore molecules. The image is assembled by deliberate translation of the zero. The zero can also be a groove. In the stochastic readout mode, a single switchable fluorophore from a random position within the diffraction zone is switched to a stable state A, while the other molecules remain in B. The coordinate is calculated from the centroid of the diffraction fluorescence spot measured by a pixelated detector. The coordinate pops up stochastically depending on where the interrogated marker molecule is located.



REversible Saturable OpticaL Fluorescence Transition s (RESOLFT)



Principle of resolution increase in a RESOLFT microscope. Horizontally a spatial dimensions is drawn. The red line corresponds to the spatially variable intensity of the light beam which drives molecules from a certain state (green) to another state (white). The blue line describes the level of saturation for this transition. Increasing the brightness (right side) of the light beam one can restrict the area, where molecules can reside in their original state (green) to arbitrarily small volumes



Rapid communication

Ground-state-depletion fluorscence microscopy: a concept for breaking the diffraction resolution limit

S.W. Hell, M. Kroug

Department of Medical Physics, University of Turku, and Centre for Biotechnology, P.O. Box 123, FIN-20521 Turku, Finland (Fax: +358-21/633-8000, E-mail: STEFAN.HELL@utu.fi)

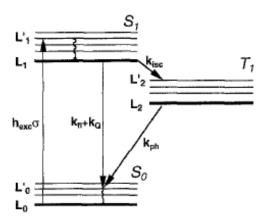


Fig. 1. Energy states of a typical dye: S_0 is the ground state, S_1 is the first singlet state, and T_1 the first triplet state

$$\begin{split} \frac{dn_0}{dt} &= -h_{\rm exc}\sigma n_0 + \left(k_{\rm fl} + k_{\rm Q}\right) n_1 + k_{\rm ph} n_2 \ , \\ \frac{dn_1}{dt} &= +h_{\rm exc}\sigma n_0 - \left(k_{\rm fl} + k_{\rm Q}\right) n_1 - k_{\rm isc} n_1 \ , \\ \frac{dn_2}{dt} &= +k_{\rm isc} n_1 - k_{\rm ph} n_2 \ , \end{split}$$

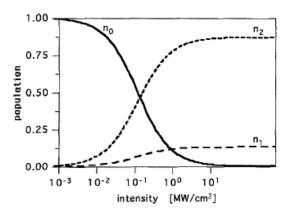


Fig. 2. The population probability of the ground state (n_0) , the first singlet state (n_1) , and the triplet state (n_2) as a function of the excitation intensity for fluoresceine

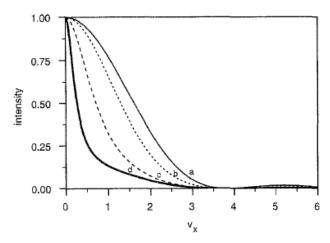


Fig. 4. Calculated effective point-spread function along the axis of the offset for peak intensities of (b) 0.01, (c) 0.1, and (d) 1 MW/cm² of the depletion beam as calculated for fluorescein, compared with the point-spread function of a classical scanning fluorescence microscope

tion, as shown in Fig. 5. A comparison of the point-spreadfunction obtained for a focal intensity of 1 MW/cm² with that of a classical scanning fluorescence microscope shows an increase in resolution by a factor of 11. Assuming an excitation wavelength of 400 nm and a numerical aperture of 1.4, one obtains a lateral FWHM of about 15 nm.

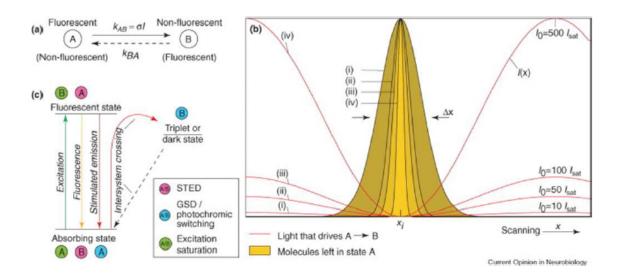
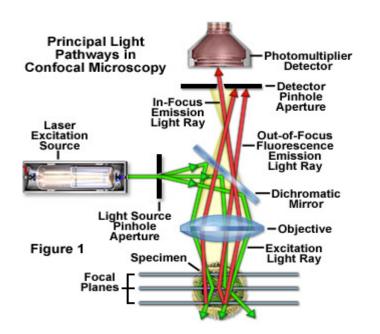
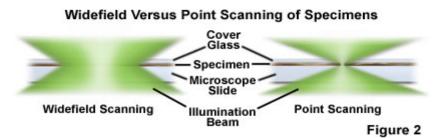


Figure 4: (Hell *et al.*, *Current Opinion in Neurobiology* 2004, 14:599–609). (a) Illustration of the fluorescence depletion principle. (b) Fluorescence density (yellow) as a function of the x-coordinate in the focal plane at STED intensities (I(x), red) with different maximal strengths (I_0). The fluorescence density (yellow) becomes sharper with increasing I_0 .

Confocal



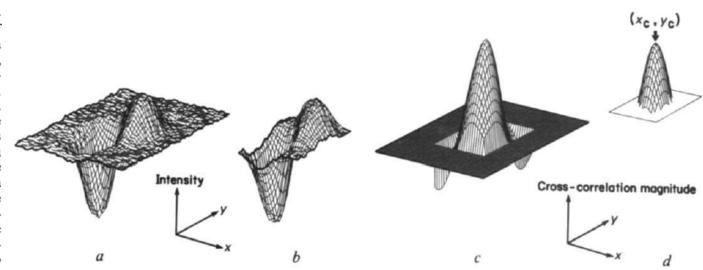


Tracking kinesin-driven movements with nanometre-scale precision

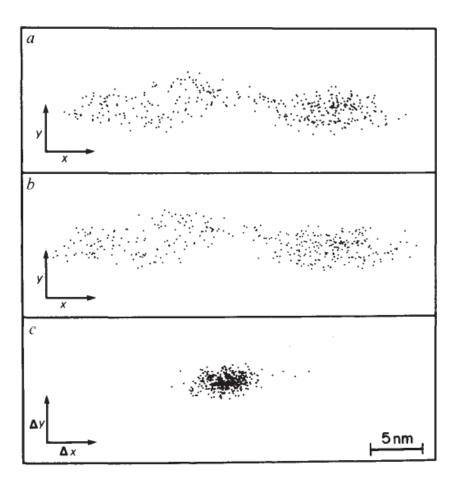
Jeff Gelles*, Bruce J. Schnapp† & Michael P. Sheetz*

NATURE VOL. 331 4 FEBRUARY 1988

Fig. 1 Stages in the highprecision measurement of relative bead position in a video frame. a, Image; b, kernel; c, cross-correlation of image and kernel; d, centroid calculation. The illustrated sequence of computations derives positional information from the entire bead image rather than from an individual point or edge only, and thereby maximizes the precision of the measurement. a, A segment of a digitized video frame containing the



video-enhanced DIC microscope image of a bead. The DIC image of a single bead consists of apposed bright and dark areas. The segment consists of a 49×63 rectangular matrix of integers, I(x, y), each of which represents the recorded light intensity at the point (x, y) in the image (x and y are integers). b, A 'kernel' segment K(x, y), consisting entirely of a single bead image. The centre of this segment is taken to be the point (0, 0). The kernel is used as a template or standard; the effect of the calculations below is to find the position in the frame segment shown in a for which the surrounding intensity distribution most closely matches that surrounding the centre of the kernel. When a sequence of consecutive video frames is analysed, a single kernel derived from one of the frames is used in the analysis of all of them. The x and y dimensions of the kernel are denoted α and β , respectively; for the case shown here, $\alpha = 27$ pixels and $\beta = 33$ pixels. c, Cross-correlation C(x, y) of the frame segment a with a scaled version of the kernel a. The cross-correlation calculation



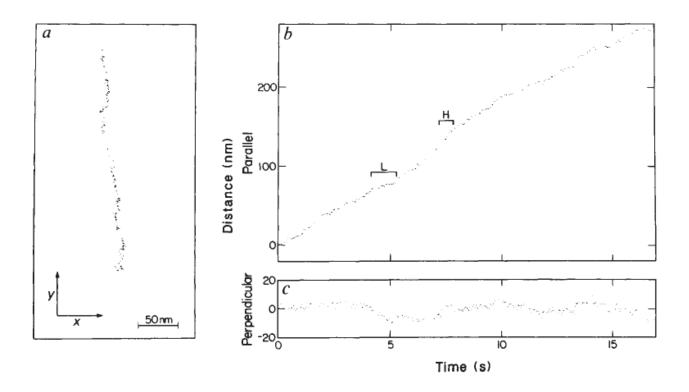
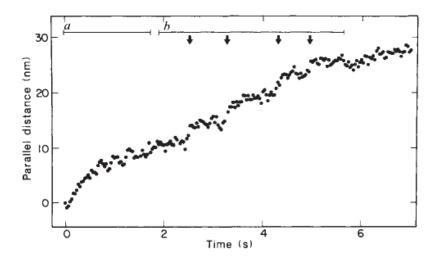


Fig. 3 Motion of a kinesin-coated bead along a microtubule in the presence of $10\,\mu M$ ATP. a, The bead

Fig. 4 Motion of a kinesin-coated bead along a microtubule in the presence of 2.5 μ M ATP. The plot shows the projection of the bead position along the microtubule axis as a function of time. The mean velocity along the microtubule axis over the period shown is 3.9 nm s⁻¹ (~0.8% of maximal velocity). Forward movement in region A occurs smoothly with a mean velocity of 5.6 nm s⁻¹. Movement in region B occurs in four discrete jumps (arrows) of length 3.7 ± 1.7 nm (mean \pm s.d.).

Methods. The sample was identical to that of Fig. 3, except for a lower ATP concentration (2.5 μM). The data-analysis procedure was identical to that of Fig. 3 except that microtubule orientation was determined manually from the positions of video cursors superimposed on the centre of the microtubule image at distances of 0.5 μm on either side of the microtubule segment on which the bead motion occurred.



Localization

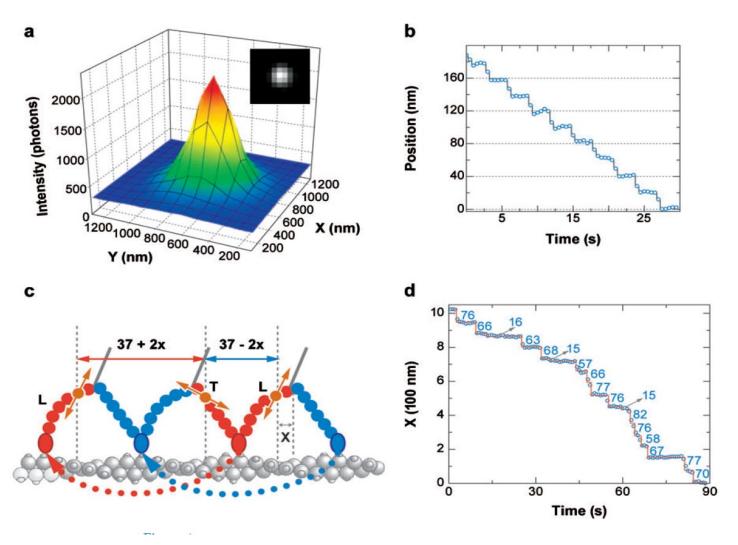
$$I(x, y) = I_{\text{background}} + A.$$

$$\exp\left\{-\frac{1}{2}\left[\left(\frac{x - x_0}{s_x}\right)^2 + \left(\frac{y - y_0}{s_y}\right)^2\right]\right\},$$

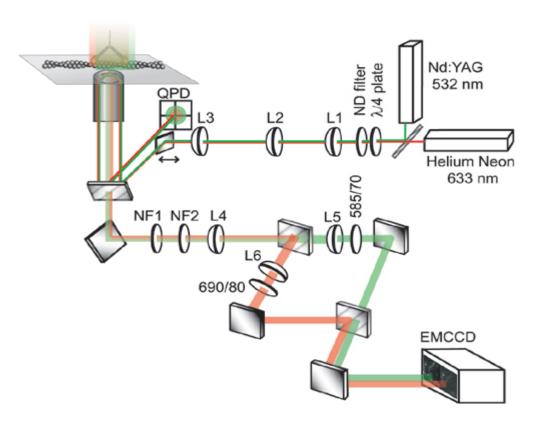
$$\sigma_{x,y}^{\mu}$$

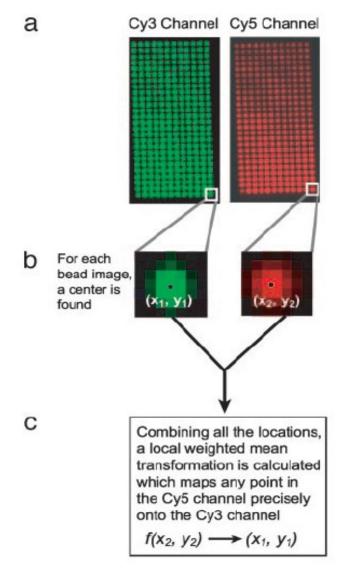
$$= \sqrt{\frac{s_{x,y}^{2}}{N_{\text{photons}}} + \frac{a^{2}}{12.N_{\text{photons}}} + \frac{8.\pi.s_{x,y}^{4}.b^{2}}{a^{2}.N_{\text{photons}}}},$$

Fluorescence Imaging with One Nanometer Accuracy (FIONA)



Fluorescence Imaging with One-Nanometer Accuracy (FIONA)





a Number of events x 10⁴ 2 10 nm 5 nm 36 nm 30 40 20 30 40 20 10 60 End-to-end distance (nm)

$$f(x, y) = z_0 + A \exp\left(\frac{1}{2}\left[\left(\frac{x - \mu_x}{\sigma_x}\right)^2 + \left(\frac{y - \mu_y}{\sigma_y}\right)^2\right]\right)$$

$$\sigma_{\mu} = \sqrt{(\sigma_{x}^{2} + \sigma_{y}^{2})} / \sqrt{N_{\gamma} - 1}.$$

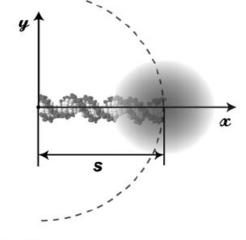
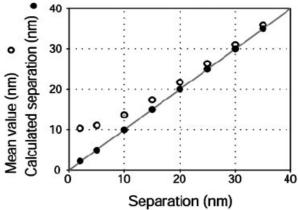


Fig. 2. Determining the accurate end-to-end distance from the skewed distribution of distance measurements. (a) The distance probability distribution was calculated by means of Monte Carlo simulations. (b) The skew in the histograms toward large values can be understood from a geometric argument. If one end of the DNA molecule is measured to reside at the origin, then only points lying on the circumference of a circle with radius s and origin (0, 0) will yield the true end-to-end distance. It is more likely that a point will lie outside of the dashed-line semicircle than inside it, which gives the distance distribution a long tail. (c) Despite the non-Gaussian nature of the distance distribution, the end-to-end separation can be calculated using the geometric mean, variance, and localization errors.



C

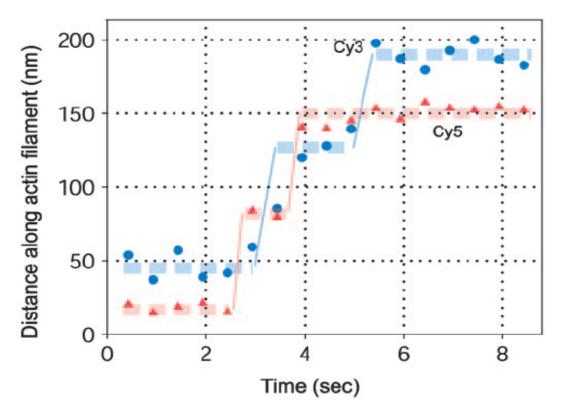
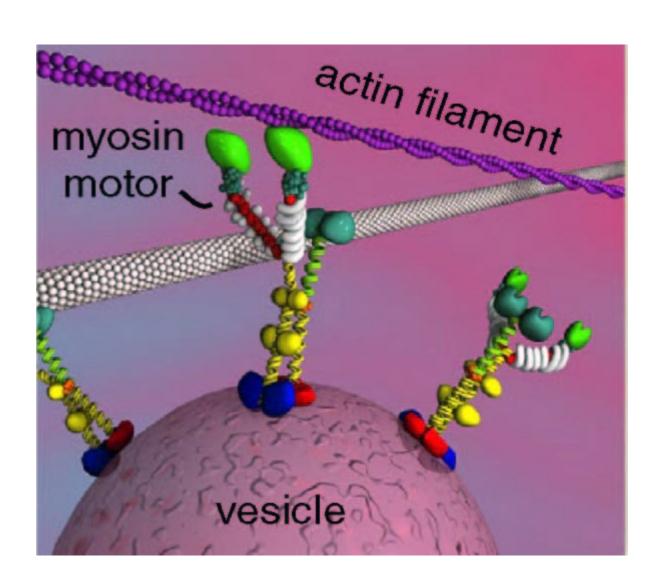
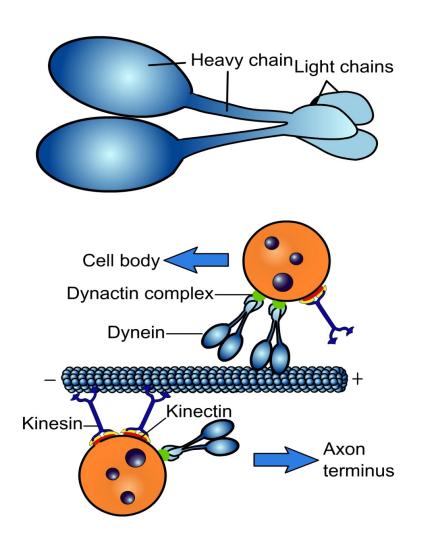


Fig. 4. Time trace of a differentially labeled myosin V molecule walking along an actin filament. The labels (Cy3 and Cy5) are covalently attached to calmodulins that were exchanged onto the myosin V molecule. In this trace, both of the fluorescent probe's locations are taking 72-nm steps, indicating that the calmodulins were exchanged close to the motor domain. The alternating positions of the probes provide a direct observation of myosin V's hand-over-hand walking mechanism.





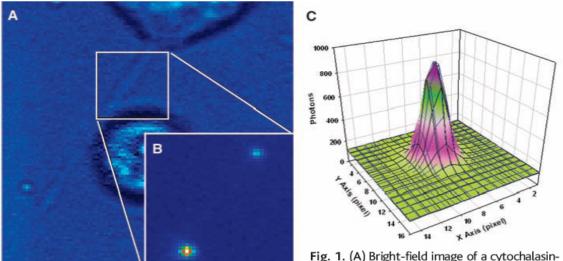
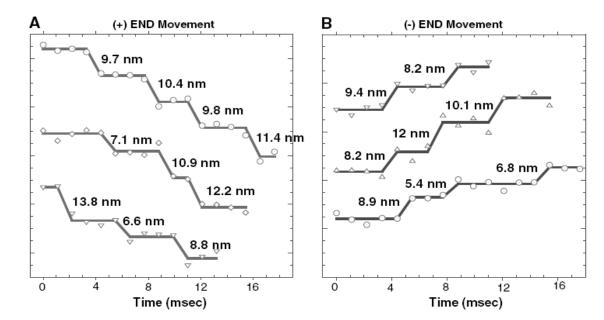


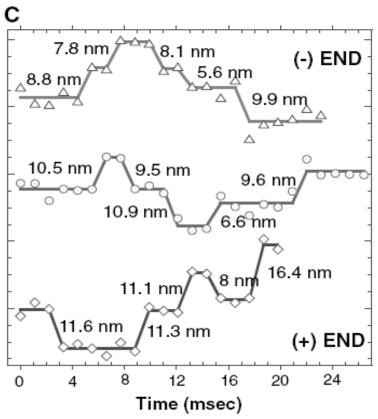
Fig. 1. (A) Bright-field image of a cytochalasin-D-treated S2 cell with a thin process. (B)

Fluorescence image of the GFP-labeled peroxisomes within the process. (C) Fluorescence image of a peroxisome can be fit to a two-dimensional Gaussian (correlation coefficient $r^2 = 0.992$), enabling the center to be determined to 1.5 nm within 1.1 ms.



Kinesin and Dynein Move a Peroxisome in Vivo: A Tug-of-War or Coordinated Movement?

SCIENCE VOL 308 3 JUNE 2005



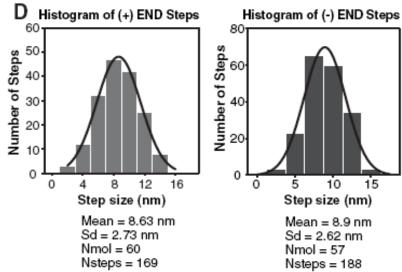
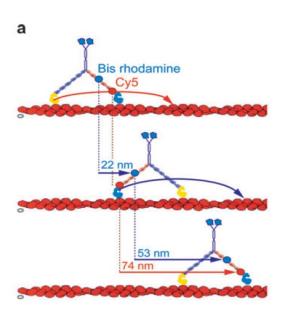
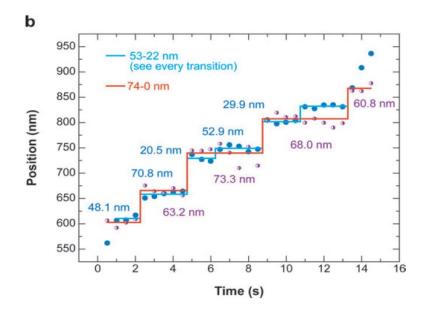


Fig. 2. Step-by-step movement of peroxisomes carried by (A) a single kinesin, (B) a single dynein, and (C) a coordination of kinesin and dynein. (D) Histograms of the individual steps of anterograde (kinesin) and retrograde (dynein) movement. A pairwise displacement of kinesin and dynein, showing the multiples of 8-nm displacement, is shown in fig. S6.

Single-molecule High-Resolution Colocalization (SHREC)





Proposed method for molecular optical imaging

E. Betzig

Here an alternative approach is proposed in which multiple discrete features within the same focal volume are spatially resolved in two steps (see Fig. 1). First, each feature is identified and isolated through one or more distinguishing optical characteristics. Second, each feature is localized; that is, its spatial coordinates are determined, on a scale that is small compared with the focal volume, either by measurement of the center of the point-spread function (PSF) associated with each feature² or by application of a spatial gradient,^{3,4} as discussed below. The complete set of coordinates for all features can then be used to reconstruct the final image in which the relative positions of the features are shown.

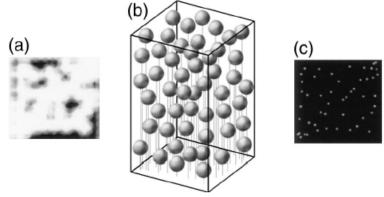
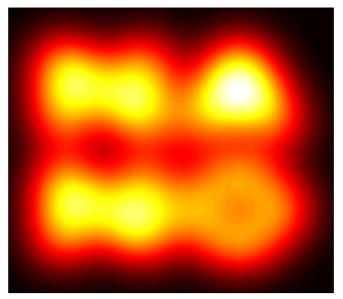
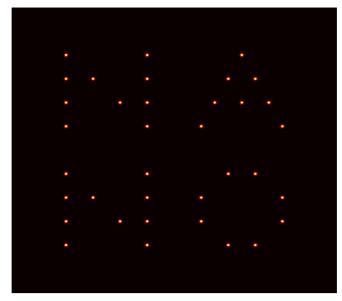


Fig. 1. (a) Field of discrete features as conventionally imaged in m spatial dimensions with a broad PSF. (b) Same features after isolation in m+n dimensions on the basis of n distinguishing optical characteristics. (c) Final image reconstructed at resolution $\delta \mathbf{x}$ given by the uncertainty in the measured position of each isolated feature. In general, $|\delta \mathbf{x}| \ll \text{PSF}$.

Photoactivation localization microscopy (PAIM)





Diffraction-limited system:

Lateral resolution $\Delta xy \approx 0.61 \, \lambda / \text{ N.A.}$ $\approx 200 \, \text{nm}$ Axis resolution $\Delta z \approx 2\lambda / \text{ N.A.}^2$ $\approx 450 \, \text{nm}$

Mean-squared position error:

$$\left(\sigma_{x,y}^{2}\right)_{m} \approx \frac{s^{2} + a^{2}/12}{N_{m}} + \frac{4\sqrt{\pi}s^{3}b_{m}^{2}}{aN_{m}^{2}}$$

s is the standard deviation of the PSF.

a is the pixel size in the image

 N_m is the total number of photons measured from molecule m b_m is the number of background photons collected in the fitting window

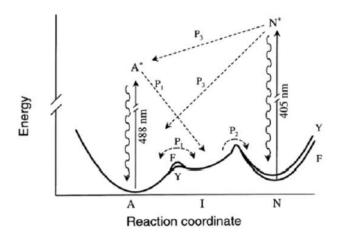
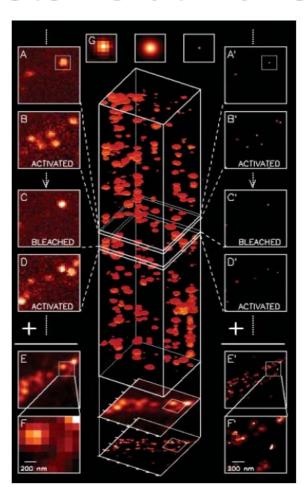


Figure 7: (Dickson *et al.*, 1997, *Nature* 388, 355-358). GFP in state A is excited to A* and returns to A upon photon emission. When I is reached from A, there is no fluorescence until I spontaneously moves to A: blinking. When I moves to N, there is no fluorescence until N is activated to N* by 405 nm excitation and GFP returns to A.

Imaging Intracellular Fluorescent Proteins at Nanometer Resolution



$$(\sigma_{x,y}^2)_m \approx \frac{s^2 + a^2/12}{N_m} + \frac{4\sqrt{\pi}s^3b_m^2}{aN_m^2}$$

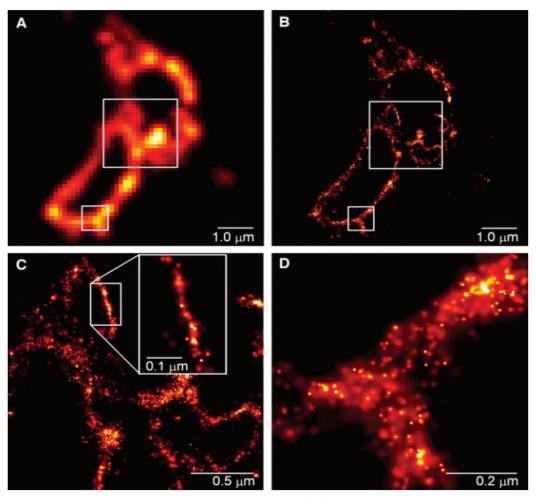


Fig. 2. Comparative summed-molecule TIRF (**A**) and PALM (**B**) images of the same region within a cryoprepared thin section from a COS-7 cell expressing the lysosomal transmembrane protein CD63 tagged with the PA-FP Kaede. The larger boxed region in (B), when viewed at higher magnification (**C**) reveals smaller associated membranes that may represent interacting lysosomes or late endosomes that are not resolvable by TIRF. In a region where the section is nearly orthogonal to the lysosomal membrane, the most highly localized molecules fall on a line of width \sim 10 nm (inset). In an obliquely cut region [(**D**), from the smaller boxed region in (B)], the distribution of CD63 within the membrane plane can be discerned.

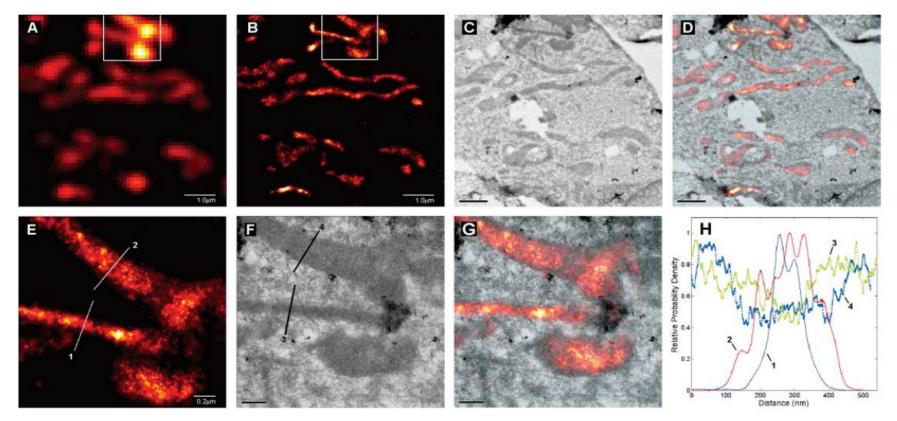


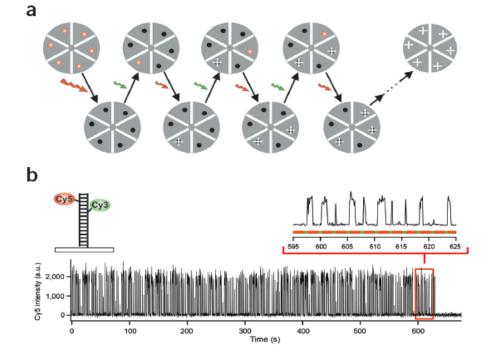
Fig. 3. Comparative summed-molecule TIRF (**A**), PALM (**B**), TEM (**C**), and PALM/TEM overlay (**D**) images of mitochondria in a cryo-prepared thin section from a COS-7 cell expressing dEosFP-tagged cytochrome-C oxidase import sequence. Higher magnification PALM (**E**), TEM (**F**), and overlay (**G**) images within the box in (B) reveal that these matrix re-

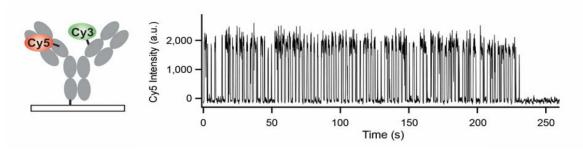
porter molecules extend up to, but not into, the $\sim\!20\text{-nm}$ outer mitochondrial membrane. The molecular distribution across two mitochondria along lines 1 and 2 in PALM image (E) are compared in (H) to the TEM signal along lines 3 and 4 in (F) across the same mitochondria. Scale bars: 1.0 μ m in (A) to (D); 0.2 μ m in (E) to (G).

Sub-diffraction-limit imaging by stochastic optical reconstruction microscopy (STORM)

Michael J Rust^{1,5}, Mark Bates^{2,5} & Xiaowei Zhuang^{1,3,4}

NATURE METHODS





: 2.2 : 1 for Cy3 and 0.1 : 1 for Cy5

Goat anti-mouse secondary antibody labeled with the cyanine switch exhibits photoswitching behavior similar to switch-labeled DNA. The antibody was labeled with Cy3 and Cy5 (as described in **Supplementary Methods** online) and bound to a quartz slide coated with unlabeled mouse anti-transferrin primary antibody. The trace shows the Cy5 fluorescence intensity detected from a single labeled antibody as it switches on and off until permanent photobleaching occurs after 230 seconds. A red laser (633 nm, 30 W/cm²) is used to excite fluorescence from Cy5 and to switch Cy5 to the dark state. A green laser (532 nm, 1 W/cm²) is used to switch the Cy5 back to the fluorescent state. The sample was excited with a sequence of alternating green and red laser pulses (0.5 s green followed by 2 s red).

$$I(x, y) = A + I_0 e^{\left[-(x'/a)^2 - (y'/b)^2\right]/2}$$

$$x' = (x - x_0)\cos\theta - (y - y_0)\sin\theta$$

$$y' = (x - x_0)\sin\theta + (y - y_0)\cos\theta$$

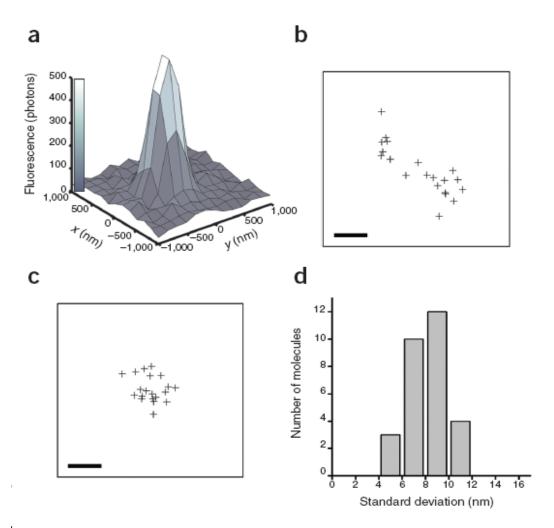
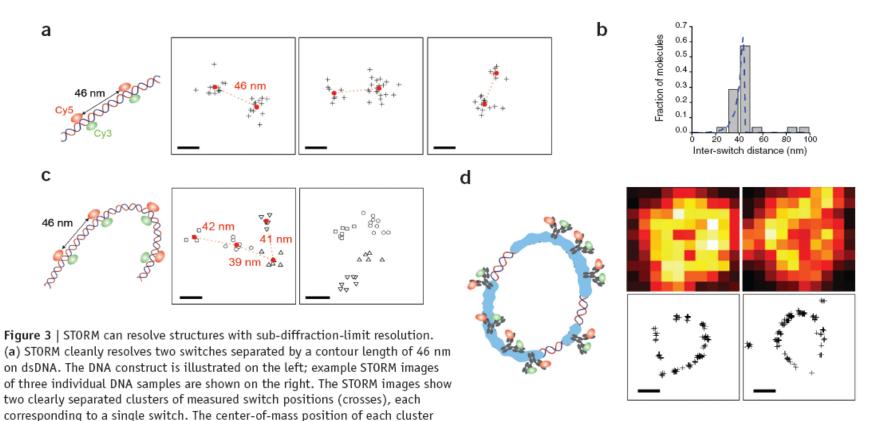


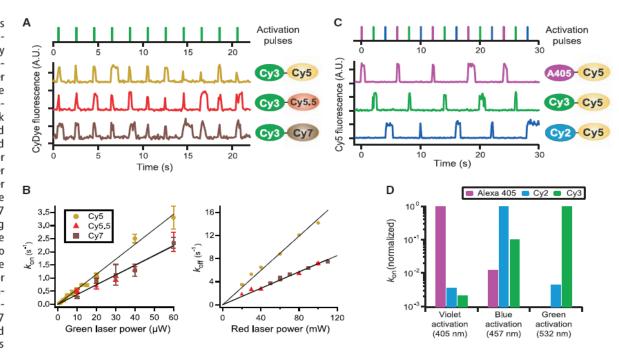
Figure 2 | The high localization accuracy of individual switches during each switching cycle defines the intrinsic resolution of STORM. (a) The point spread function (PSF) of the emission from a single switch on DNA during a single switching cycle. Fitting the PSF to a two-dimensional Gaussian (not shown) gives the centroid position of the PSF. (b,c) The centroid positions of an individual switch determined in 20 successive imaging cycles before (b) and after (c) correction for sample drift. Scale bars, 20 nm. (d) A histogram of the standard deviation of centroid positions. The standard deviation is determined as $(\sigma_x + \sigma_y) / 2$ for each switch using 20 imaging cycles, where σ_x and σ_y are the standard deviations of the centroid positions in the x and y dimensions. This histogram was constructed from 29 switches.



is marked by a red dot. The inter-switch distances are 46 nm, 44 nm and 34 nm for these three examples. Scale bars, 20 nm. (b) Comparison between the inter-switch distances measured using STORM (columns) and the predicted distance distribution considering the flexibility of DNA (dashed line). (c) STORM images of four switches attached to a dsDNA, pair-wise separated by a contour length of 46 nm. The measured switch positions are clustered by an automated algorithm and different clusters are indicated by different symbols. Scale bars, 20 nm. (d) STORM images of RecA-coated circular plasmid DNA. Indirect immunofluorescence images with switch-labeled secondary antibody taken by a total internal reflection microscope (top); the reconstructed STORM images of the same filaments (bottom). Scale bars, 300 nm.

Multicolor Super-Resolution Imaging with Photo-Switchable Fluorescent Probes SCIENCE VOL 317 21 SEPTEMBER 2007

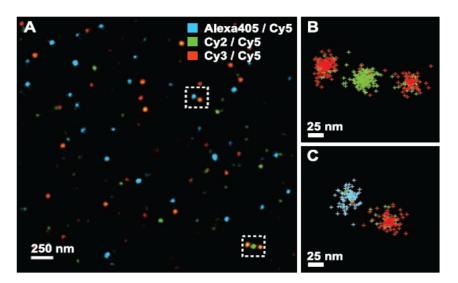
Fig. 1. Photo-switchable probes constructed from activatorreporter pairs. (A) Spectrally distinct reporters exhibit photoswitching behavior. The lower panel shows the fluorescence time traces of three photoswitchable reporters Cv5 (dark yellow line), Cy5.5 (red line), and Cy7 (brown line), each paired with a Cv3 dve as the activator on a DNA construct. The upper panel shows the green laser pulses (532 nm) used to activate the reporters. The red laser (657) nm) was continuously on, serving to excite fluorescence from the reporters and to switch them to the dark state. Traces were shifted relative to each other for clarity. (B) Switching rate constants k_{on} and k_{off} of the Cy3-Cy5, Cy3-Cy5.5, and Cy3-Cy7 pairs as a function of green and red laser power. Error bars

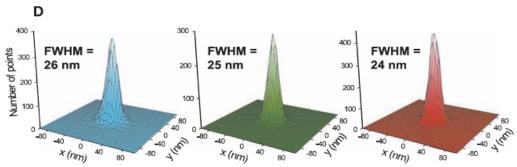


indicate SEM from about three data sets. The laser power to intensity calibration may vary between different samples because of moderate differences in the laser spot size at the sample. (C) The same reporter can be activated by spectrally distinct activators. The lower panel shows the fluorescence time traces of Cy5 paired with three different activators, Alexa 405 (magenta line), Cy2 (blue line), and Cy3 (green line). The upper panel shows the violet (405 nm, magenta line), blue (457 nm, blue line), and green (532 nm, green line) activation pulses. (D)

Normalized activation rate constants of the three dye pairs at three activation wavelengths: 405, 457, and 532 nm. The $k_{\rm cn}$ values of Alexa 405-Cy5, Cy2-Cy5, and Cy3-Cy5 were used for normalization at 405, 457, and 532 nm, respectively. The absolute activation rates were rapid for each pair at its corresponding optimal wavelength, with values of 10 s⁻¹ or greater at only a few hundred μ W of laser power. The activation rate of the Alexa 405-Cy5 pair by the 532-nm laser was too small to be measured reliably.

Fig. 2. Three-color STORM imaging of a model DNA sample. (A) Three-color STORM image of three different DNA constructs labeled with Alexa 405-Cy5, Cy2-Cy5, or Cy3-Cy5 mixed at a high surface density on a microscope slide. The image was plotted by rendering each localization as a Gaussian peak, the width of which was scaled with the theoretical localization accuracy given by the number of photons detected (26). Each colored spot in this image represents a cluster of localizations from a single DNA molecule. A conventional fluorescence image of the same area is shown in fig. S3 for comparison. (B and **C**) Higher-magnification views of the boxed regions in (A) show several examples of closely spaced DNA molecules. Each localization was plotted as a cross, colored according to the following code: If the molecule was activated by a 405-, 457-, or 532-nm laser pulse, the color of the cross was assigned as blue, green, or red, respectively. (D) The localization distributions of the blue, green, or red clusters. The two-dimensional histograms of localizations were generated by aligning multiple (50 to 60) clusters by their center of mass. The histograms were fit to a Gaussian profile to determine their FWHM.





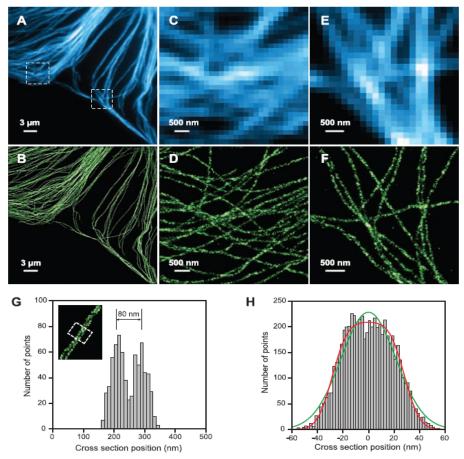


Fig. 3. STORM imaging of microtubules in a mammalian cell. **(A)** Conventional immunofluorescence image of microtubules in a large area of a BS-C-1 cell. **(B)** STORM image of the same area. **(C** and **E)** Conventional and **(D** and **F)** STORM images corresponding to the boxed regions in **(A)**. **(G)** Cross-sectional profiles of two nearby microtubule filaments in the cell. The inset shows the STORM image, and the histogram shows the cross-sectional distribution of localizations within the small regions specified by the white box. **(H)** Cross-sectional profile of a microtubule segment determined from the STORM image. A relatively long segment (\sim 7 μ m) was chosen to obtain good statistics. The histogram shows the cross-sectional distribution of localizations. The green line is a single Gaussian fit with FWHM = 51 nm. The red line shows the fit obtained by convolving a rectangular function of width = d with a Gaussian function of FWHM = r. The fit yields d = 56 nm and r = 22 nm, corresponding to the antibody-coated microtubule width and the imaging resolution, respectively.

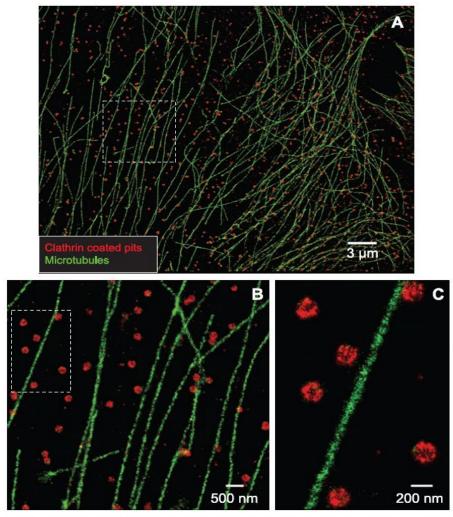


Fig. 4. Two-color STORM imaging of microtubules and CCPs in a mammalian cell. **(A)** STORM image of a large area of a BS-C-1 cell. The secondary antibodies used for microtubule staining were labeled with Cy2 and Alexa 647, and those for clathrin were labeled with Cy3 and Alexa 647. The 457- and 532-nm laser pulses were used to selectively activate the two pairs. Each localization was false colored according to the following code: green for 457-nm activation and red for 532-nm activation. **(B)** STORM image corresponding to the boxed region in (A) shown at a higher magnification. **(C)** Further magnified view of the boxed region in (B).

Three-Dimensional Super-Resolution Imaging by Stochastic Optical Reconstruction Microscopy

8 FEBRUARY 2008 VOL 319 SCIENCE

Fig. 1. The scheme of 3D STORM. (A) Threedimensional localization of individual fluorophores. The simplified optical diagram illustrates the principle of determining the z coordinate of a fluorescent object from the ellipticity of its image by introducing a cylindrical lens into the imaging path. The right panel shows images of a fluorophore at various z positions. EMCCD, electronmultiplying charge-coupled device. (B) Calibration curve of image widths w_x and w_y as a function of z obtained from single Alexa 647 molecules. Each data point represents the average value obtained from six molecules. The data were fit to a defocusing function (red curve) as described in (27). (C) Three-dimensional localization distribution of single molecules. Each molecule gives a cluster of localizations due to repetitive activation of the same molecule. Localizations from 145 clusters were aligned by their center of mass to generate the overall 3D presentation of the localization distribution (left panel). Histograms of the distribution in x, y, and z (right panels) were fit to a Gaussian function, yielding standard deviations of 9 nm in x, 11 nm in y, and 22 nm in z.

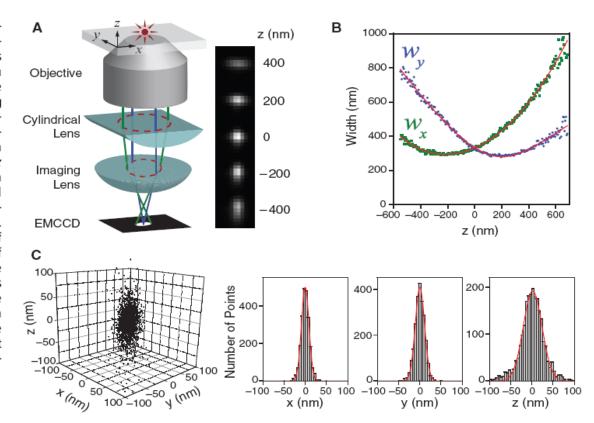
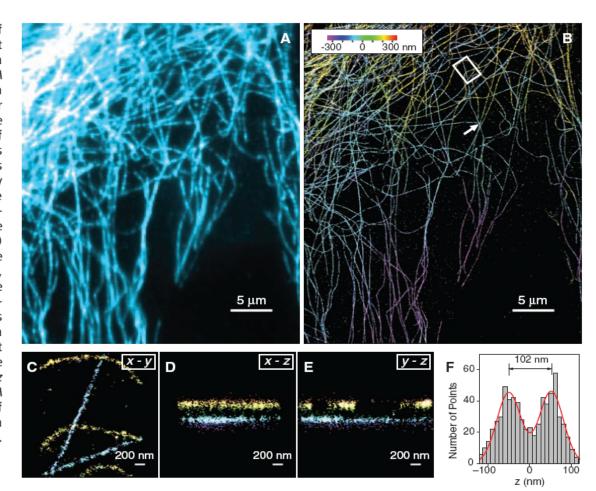


Fig. 2. Three-dimensional STORM imaging of microtubules in a cell. (A) Conventional indirect immunofluorescence image of microtubules in a large area of a BS-C-1 cell. (B) The 3D STORM image of the same area, with the z-position information color-coded according to the color scale bar. Each localization is depicted in the STORM image as a Gaussian peak, the width of which is determined by the number of photons detected (5). (**C** to **E**) The x-y, x-z, and y-z cross sections of a small region of the cell outlined by the white box in (B), showing five microtubule filaments. Movie S1 shows the 3D representation of this region, with the viewing angle rotated to show different perspectives (27). (F) The z profile of two microtubules crossing in the x-y projection but separated by 102 nm in z, from a region indicated by the arrow in (B). The histogram shows the distribution of z coordinates of the localizations, fit to two Gaussians with identical widths (FWHM = 66 nm) and a separation of 102 nm (red curve). The apparent width of 66 nm agrees quantitatively with the convolution of our imaging resolution in z (represented by a Gaussian function with FWHM of 55 nm) and the previously measured width of antibody-coated microtubules (represented by a uniform distribution with a width of 56 nm) (5).



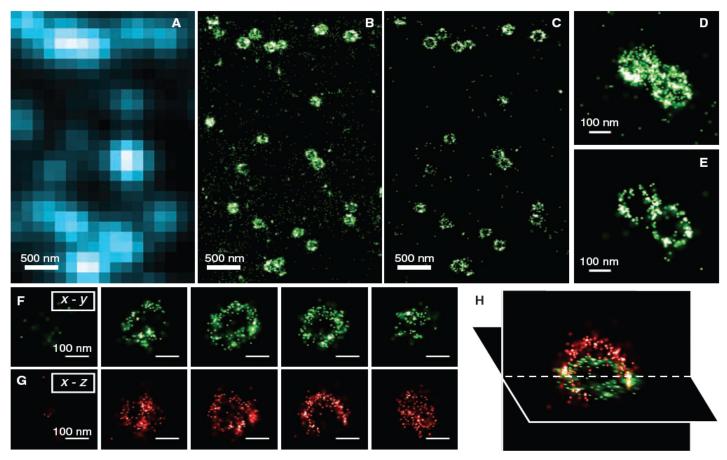
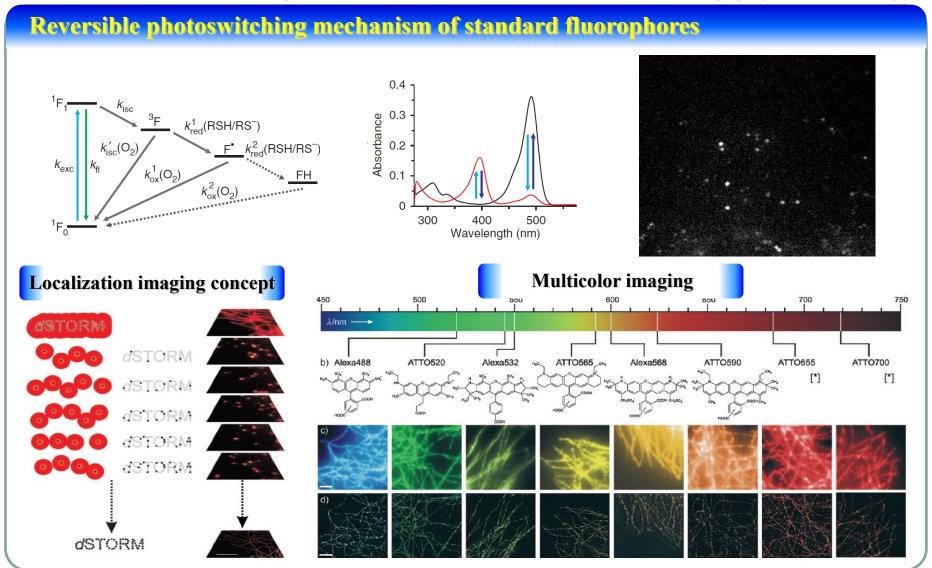


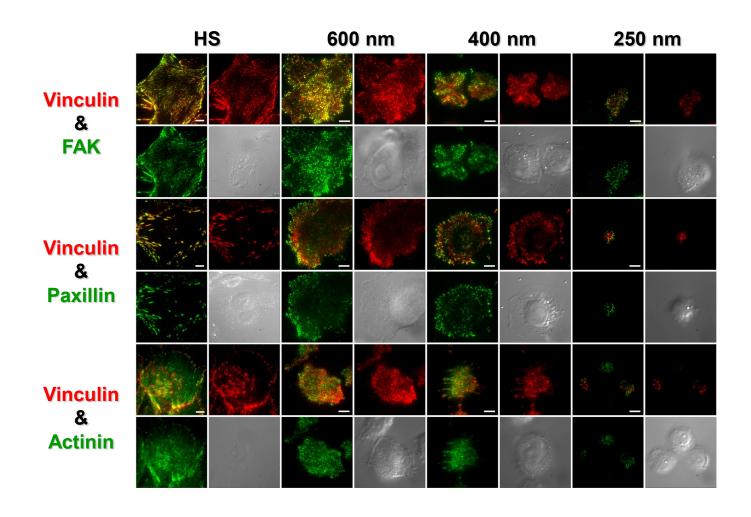
Fig. 3. Three-dimensional STORM imaging of clathrin-coated pits in a cell. **(A)** Conventional direct immunofluorescence image of clathrin in a region of a BS-C-1 cell. **(B)** The 2D STORM image of the same area, with all localizations at different *z* positions included. **(C)** An *x-y* cross section (50 nm thick in *z*) of the same area, showing the ring-like structure of the periphery of the CCPs at the

plasma membrane. (**D** and **E**) Magnified view of two nearby CCPs in 2D STORM (D) and their *x-y* cross section (100 nm thick) in the 3D image (E). (**F** to **H**) Serial *x-y* cross sections (each 50 nm thick in *z*) (F) and *x-z* cross sections (each 50 nm thick in *y*) (G) of a CCP, and an *x-y* and *x-z* cross section presented in 3D perspective (H), showing the half-spherical cage-like structure of the pit.

Direct stochastic optical reconstruction microscopy (dSTORM)



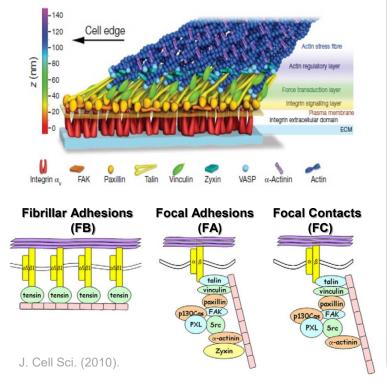
Focal adhesion complexes on HS and pattern ECM surface

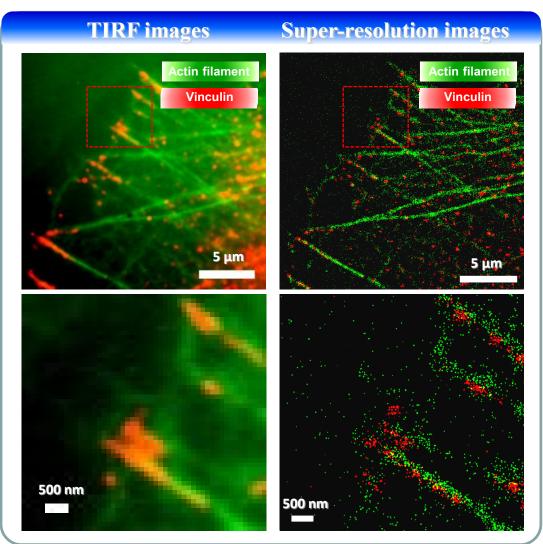


Multicolor superresolution fluorescence imaging

Objective

- 1. Characteristic distance between focal adhesion complexes.
- 2. Cross-correlation between different focal adhesion complexes.
- 3. Number of focal adhesion complexes at different adhesion types.



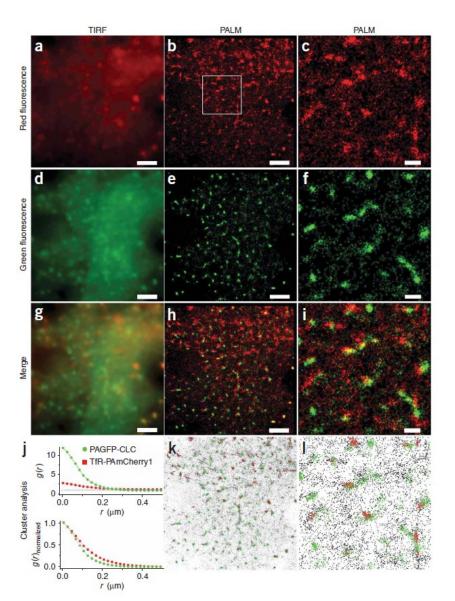


Pair correlation analysis

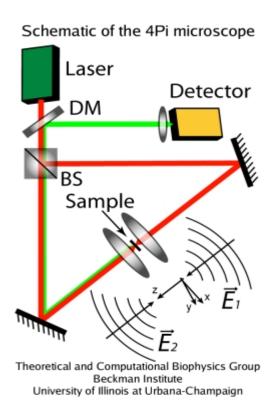
$$g(r) = \left\langle \frac{\rho(r)}{\overline{\rho}} \right\rangle$$

 $\rho(r)$ is the density of molecules at distance. r from a given molecule.

 $\overline{\rho}$ is the average density of molecules, with this ratio averaged over the entire ensemble of molecules.



4 Pi



Theoretical and Computational Biophysics Group

Beckman Institute

University of Illinois at Urbana-Champaign

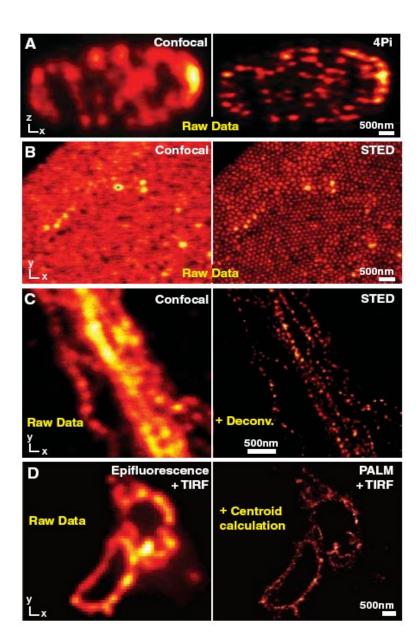
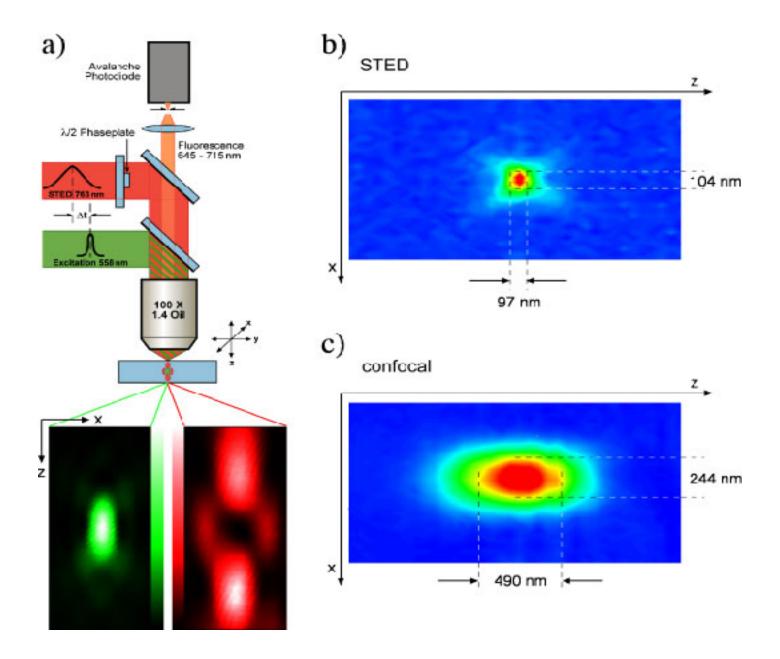
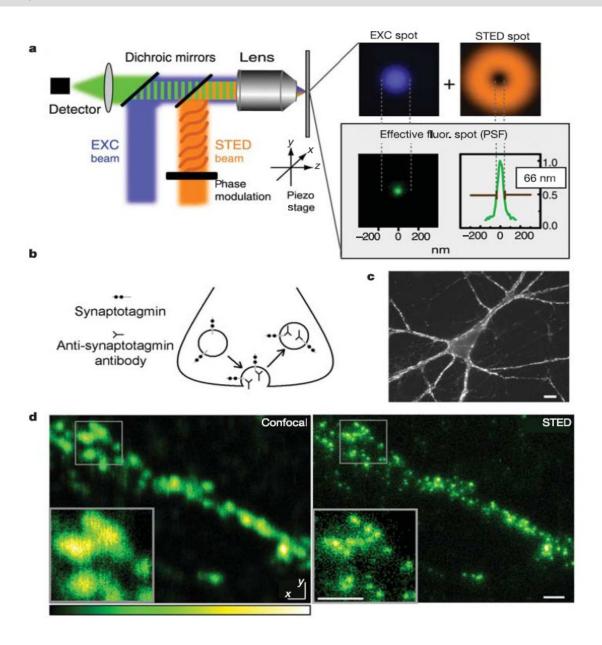
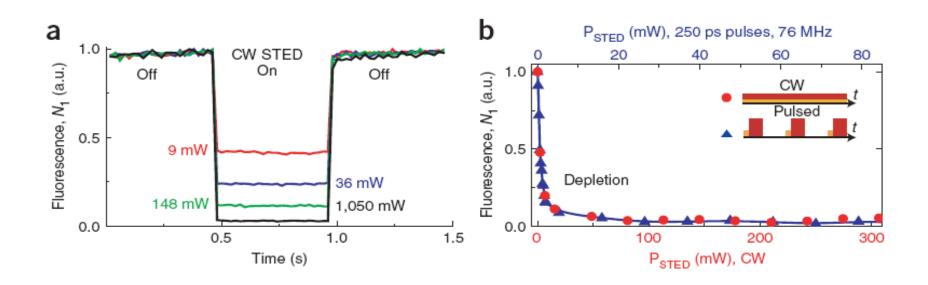


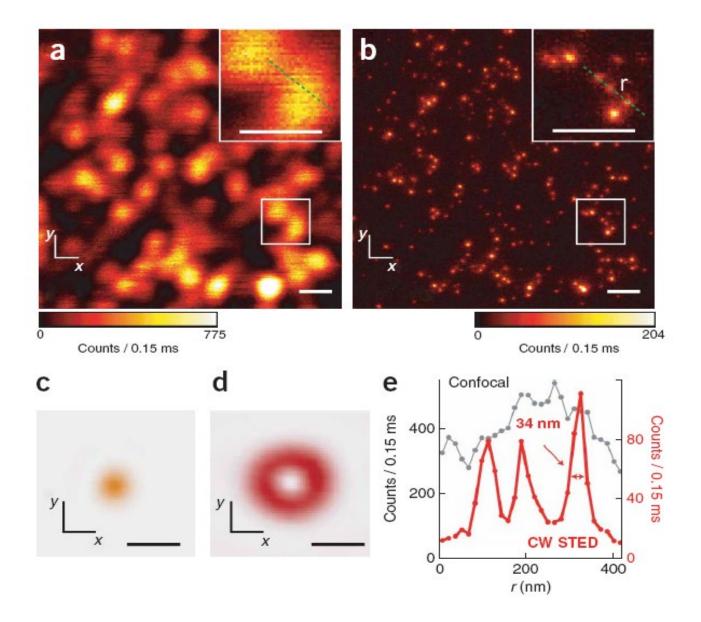
Fig. 4. Side-by-side comparisons. **(A)** Confocal versus 4Pi axial (xz) image of microtubules in a neuron: 4Pi image displays 140-nm z resolution; lens of $\alpha=74^{\circ}$ and with two-photon excitation at 800 nm. The plain 4Pi image is due to a narrow solitary peak without lobes; mathematical lobe-removal is not required. **(B)** Unlike the confocal reference, the STED image reveals the spatial order of self-assembled fused silica nanobeads containing a fluorescence core (45). **(C)** Neurofilaments in human neuroblastoma recorded in the confocal mode (left) and with STED after nonlinear deconvolution (right) displaying a focal plane resolution of 20 to 30 nm (39). **(D)** Epifluorescence versus PALM recording of a cryoprepared section from a mammalian cell expressing a lysosomal transmembrane protein tagged with a photoswitchable protein; both images were recorded with a TIRF setup. PALM resolution ranges between 20 and 60 nm, whereas individual protein localizations can be 2 nm (12).

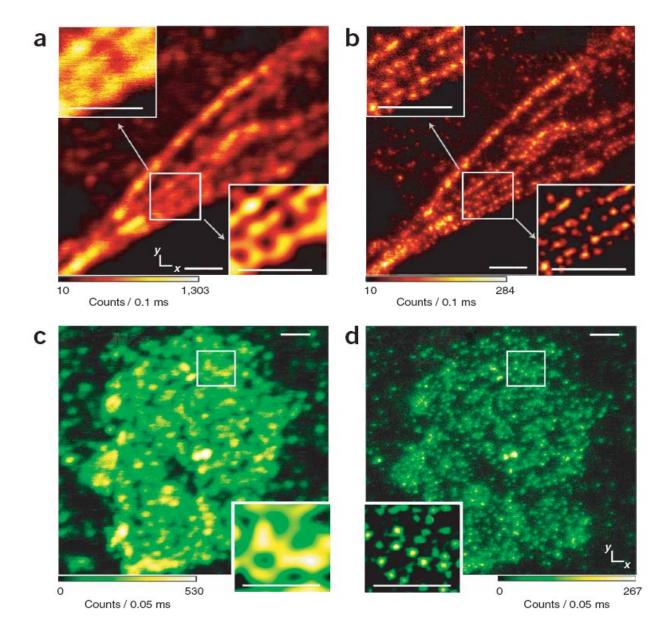




STED microscopy with continuous wave beams







SHORT COMMUNICATION

Surpassing the lateral resolution limit by a factor of two using structured illumination microscopy

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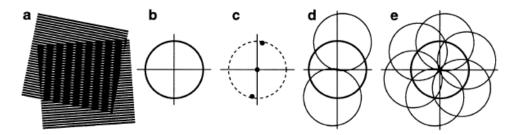


Fig. 1. Concept of resolution enhancement by structured illumination. (a) If two line patterns are superposed (multiplied), their product will contain moiré fringes (seen here as the apparent vertical stripes in the overlap region). (b) A conventional microscope is limited by diffraction. The set of low-resolution information that it can detect defines a circular 'observable region' of reciprocal space. (c) A sinusoidally striped illumination pattern has only three Fourier components. The possible positions of the two side components are limited by the same circle that defines the observable region (dashed). If the sample is illuminated with such structured light, moiré fringes will appear which represent information that has changed position in reciprocal space. The amounts of that movement correspond to the three Fourier components of the illumination. The observable region will thus contain, in addition to the normal information, moved information that originates in two offset regions (d). From a sequence of such images with different orientation and phase of the pattern, it is possible to recover information from an area twice the size of the normally observable region, corresponding to twice the normal resolution (e).

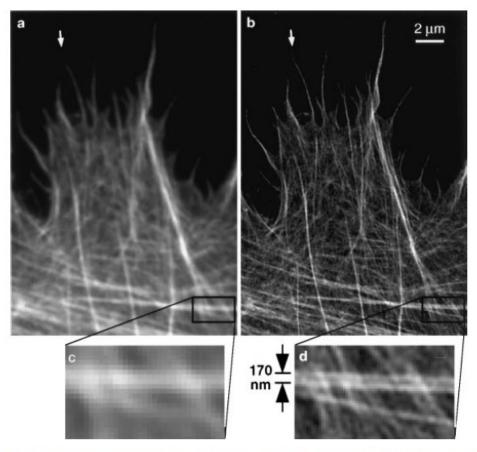
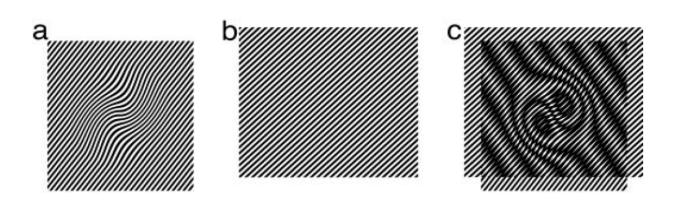


Fig. 4. The actin cytoskeleton at the edge of a HeLa cell, as imaged by (a, c) conventional and (b, d) structured illumination microscopy. (c, d) Enlargements of the boxed areas in (a) and (b), respectively. Fibres separated by less than the resolution limit of the conventional microscope are well resolved using structured illumination (d). The apparent widths (FWHM) of the finest protruding fibres [small arrows in (a, b)] are lowered to 110-120 nm in (b), compared to 280-300 nm in (a).

Nonlinear structured-illumination microscopy: Wide-field fluorescence imaging with theoretically unlimited resolution



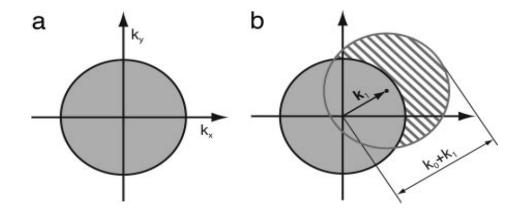
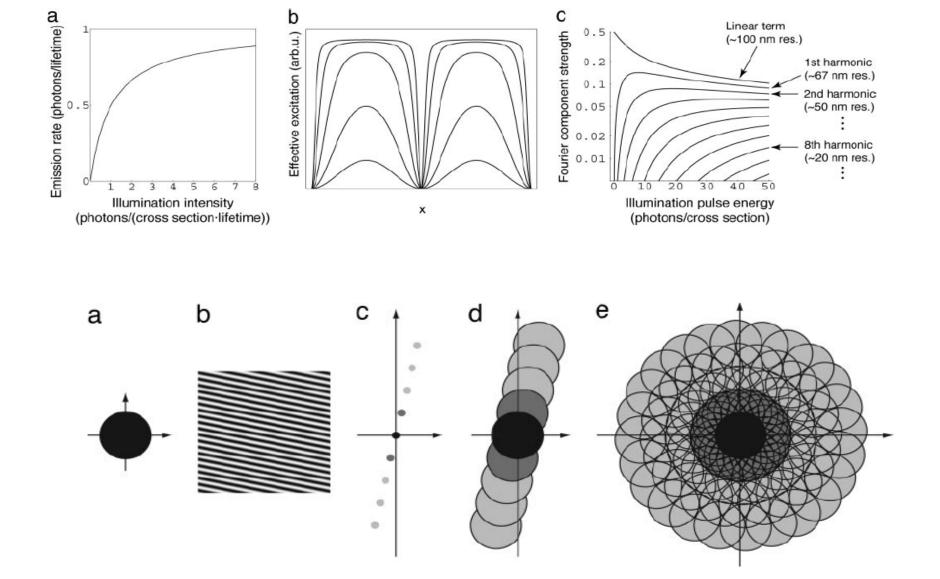


Fig. 2. Structured-illumination concept. (a) The set of sample spatial frequencies that can be observed by the conventional microscope defines a circular observable region of radius k_0 in frequency space. (b) If the excitation light contains a spatial frequency k_1 , a new set of information becomes visible in the form of moiré fringes (hatched circle). This region has the same shape as the normal observable region but is centered at k_1 . The maximum spatial frequency that can be detected (in this direction) is $k_0 + k_1$.

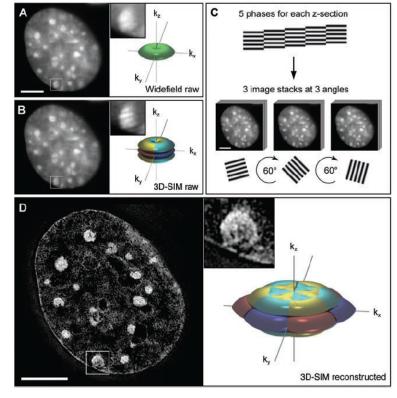


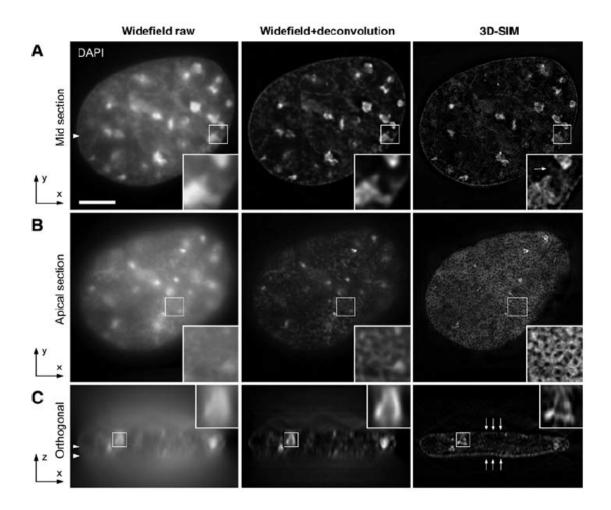
Subdiffraction Multicolor Imaging of the Nuclear Periphery with 3D Structured Illumination Microscopy

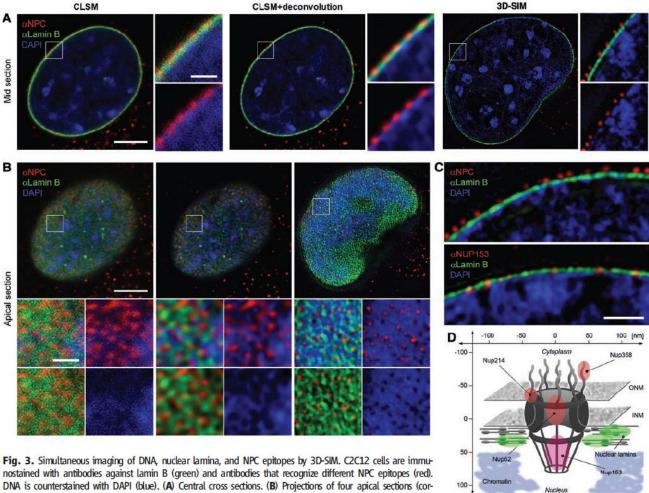
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> Fig. 1. Subdiffraction resolution imaging with 3D-SIM. (A and B) Cross section through a DAPI-stained C2C12 cell nucleus acquired with conventional wide-field illumination (A) and with structured illumination (B), showing the striped interference pattern (inset). The renderings to the right illustrate the respective support of detection in frequency space. The axes k_x , k_y , and k_z indicate spatial frequencies along the x, y, and z directions. The surfaces of the renderings represent the corresponding resolution limit. The depression of the frequency support ("missing cone") in z direction in (A) indicates the restriction in axial resolution of conventional wide-field microscopy. With 3D-SIM, the axial support is extended but remains within the resolution limit. (C) Five phases of the sine wave pattern are recorded at each z position, allowing the shifted components to be separated and returned to their proper location in frequency space. Three image stacks are recorded with the diffraction grating sequentially rotated into three positions 60° apart, resulting in nearly rotationally symmetric support over a larger region of frequency space. (D) The same cross section of the reconstructed 3D-SIM image shows enhanced image details compared with the original image (insets). The increase in resolution is shown in frequency space on the right, with the coverage extending two times farther from the origin. Scale bars indicate 5 µm.







responding to a thickness of 0.5 μm). Boxed regions are shown below at 4× magnification; scale bars indicate

5 µm and 1 µm, respectively. (A) CLSM and deconvolution still show partially overlapping signals. In contrast, with 3D-SIM the spatial separation of NPC, lamina, and chromatin and chromatin-free channels underneath nuclear pores are clearly visible. (B) Top view on the nuclear envelope. Whereas CLSM fails to resolve close nuclear pores, 3D-SIM shows clearly separated NPC signals at voids of peripheral chromatin and surrounded by an irregular network of nuclear lamina. (C) Mid sections comparing stainings with an antibody that mainly reacts with Nup214, Nup358, and Nup62 (α NPC) and one specifically recognizing Nup153 (α Nup153). The α NPC signal is above the lamina (140 \pm 8 nm), whereas the α Nup153 pore signal is at the same level as the lamina (-15 \pm 20 nm). Scale bars 1 μ m. (**D**) Schematic outline of the NPC, showing the relative position of Nup proteins and surrounding structures. ONM, outer nuclear membrane; INM, inner nuclear membrane.

Super-resolution video microscopy of live cells by structured illumination

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Structured-illumination microscopy can double the resolution of the widefield fluorescence microscope but has previously been too slow for dynamic live imaging. Here we demonstrate a high-speed structured-illumination microscope that is capable of 100-nm resolution at frame rates up to 11 Hz for several hundred time points. We demonstrate the microscope by video imaging of tubulin and kinesin dynamics in living *Drosophila melanogaster* S2 cells in the total internal reflection mode.

