



Microfluidics – „Lab on a Chip“

Outline

- introduction to microfluidics
- physics of micro-scale
- lab on a chip applications
 - life and medical science
 - discovery of novel proteins
 - protein and metabolic engineering
- design and fabrication
- sensing and detection

Bi7430 Molecular Biotechnology

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Lab on a Chip Concept



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Microfluidics

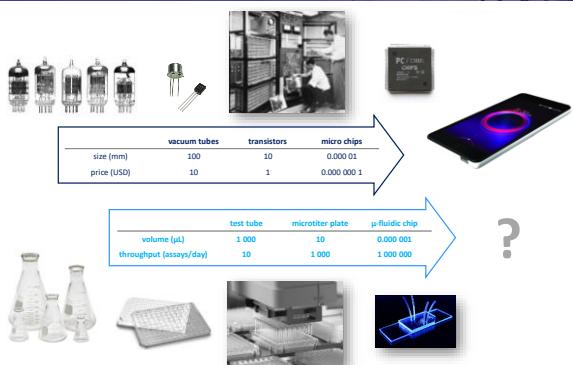
- „behavior, control and manipulation of fluids geometrically constrained to a small dimensions“
 - dimensions ($1'$ - $100'$ μm)
 - volumes (nL , pL , fL)
 - unrivalled precision of control
 - (ultra)high analytical throughput
 - reduced sample and power consumption
 - facile process integration and automation



Nature 507, 181 (2014)

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Miniaturization & integration



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Concepts in microfluidics

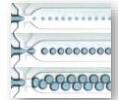
continuous-flow microfluidics

manipulation of continuous liquid flow through micro-fabricated channels



droplet-based microfluidics

manipulating discrete volumes of fluids in immiscible phases



digital microfluidics

droplets manipulated on a substrate using electro-wetting

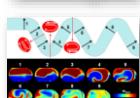
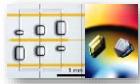


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Novel Physics of Micro-Scale

viscosity, surface tension and capillary forces dominate

- lack of turbulent phenomena
 - + nontrivial chemical gradients to study chemotaxis
- absence of density-driven convection
 - + free interface diffusion, efficient protein crystallization kinetics
- strong shearing forces
 - + fast mixing kinetics of protein folding and/or catalysis



Nature Biotechnol., 20, 826 (2002) *Appl. Phys. Lett.*, 83, 4664 (2003) *PNAS* 99, 16531 (2002)

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Lab on a Chip applications

analytics and chemistry



PCR and sequencing



point of care diagnostics

pharmacology



clinical studies

single cell biology



high throughput biology

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Polymerase chain reaction



□ classical PCR

- slow heating/cooling cycles
- PCR tubes (strips), 96-well MTP
- volume 50 to 500 µL



Kary Mullis
Nobel Prize in 1993



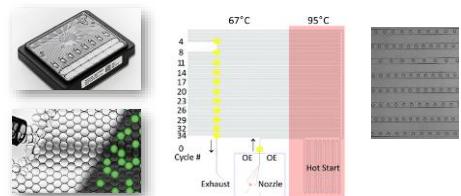
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Digital polymerase chain reaction



□ digital PCR (single molecule)

- 1 nanoliter droplets
- 20 000 droplets per run
- fast heat transfer



Analytical Chemistry, 2008, 80, 8975

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Next-generation sequencing



□ parallelization of single molecule pyrosequencing

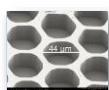
□ 454 Pyrosequencing (Roche)

water in oil droplets 1 picoliter (10^{-12} liters)

1 mil. reads/run, 10 USD/Mbase

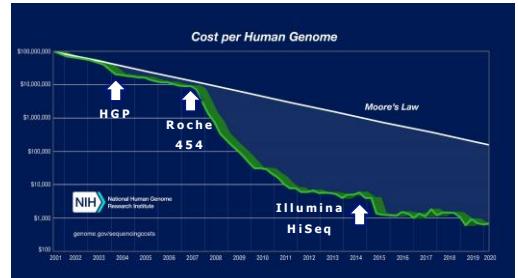


Frederick Sanger
Nobel Prize in 1980



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Revolution in DNA analysis



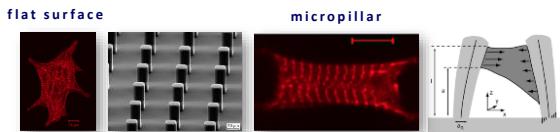
□ 2003: 13 years, 3 billion USD

□ 2018: days, < 1,000 USD

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Organ(oid)s on chip

- ❑ 3D chips mimicking human's physiological responses
(e.g., pathological, pharmacokinetic, toxicological)
- ❑ realistic *in vitro* model closer to *in vivo* cell environment
(e.g., mechanical strain, patterning, fluid shear stresses)

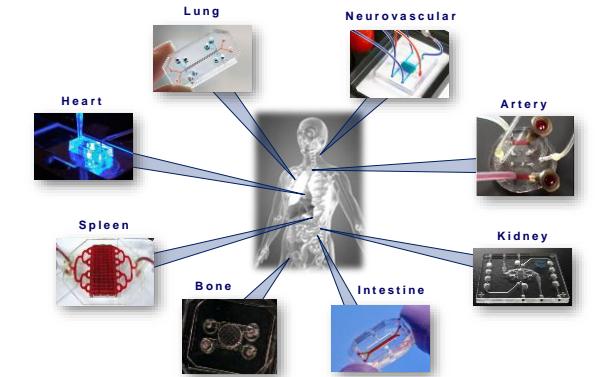


Nature 471, 661–665 (2011)

Biophysical Journal 94(5) 1854–1866

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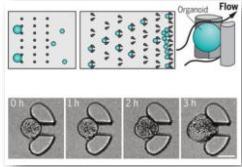
Organs on chip



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Organ(oid)s on chip

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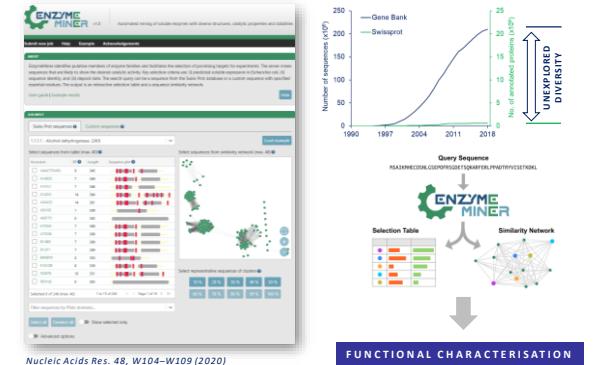
- ❑ can replace expensive and controversial animal testing

Nature 471, 661–665 (2011)

Science 364, 960–965 (2019)

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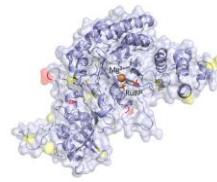
Protein Discovery and Engineering



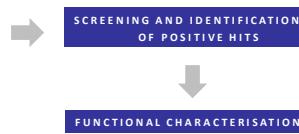
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Protein Discovery and Engineering

1-WQSEKTYVNLALKEELIAGGRRVLCATIMKPKAYQYVATAAHFAKESG-59
 51-TOTIVVECYTDDFTG-DALUYVCEBARLTKIAFYVALVDFRITIGGM-106
 101-IAIPFLSTLTHNQHQGSDVEYAKHOFVYRAYPAFLPQFVNPIDALMRYL-150
 151-GREPVDTLGLGVSTKPLKLGSRPFPAEKAQNGGFTNDRPQFVNPIDALMRYL-200
 201-FAFSRQDIAVADAMPAQEESEKAPLFANITAOFPFIAAEGYLET-250
 251-FGEKAHSLVALLDVTYNGAATTTAKRPRDFNLFTRHAGNGAVTQGQK-300
 301-KOTTAPEVKCRAALDQSGKHTHTMFGFQHGRGKSRATAXMTQEAQD-350
 351-PFPTRQHGQHGAKCTTISQGMDLAFYFENGLQHVVLTAGGGAGFH-400
 401-DOVYADASLSLAQWQHNDQVPI-DYAREKEELAAAPFEPFLGDAQ-YFO-450
 451-WKQALDQFDTTSALELP-466

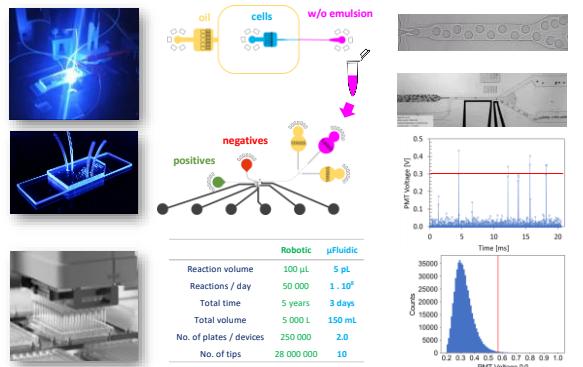


No.	Coverage (95%)
1	94
2	3 066
3	98 163
4	3 141 251
5	100 520 093



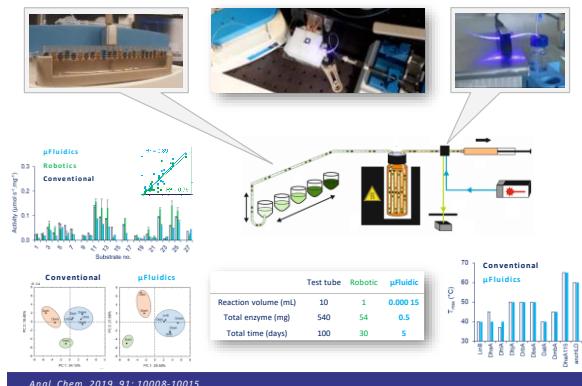
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High-throughput on chip sorting (FADS)



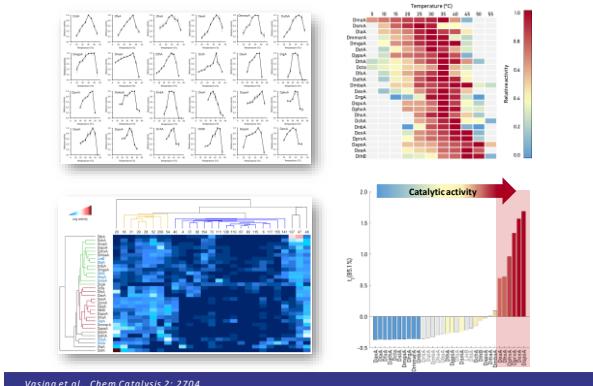
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Activity and specificity in μ-droplets

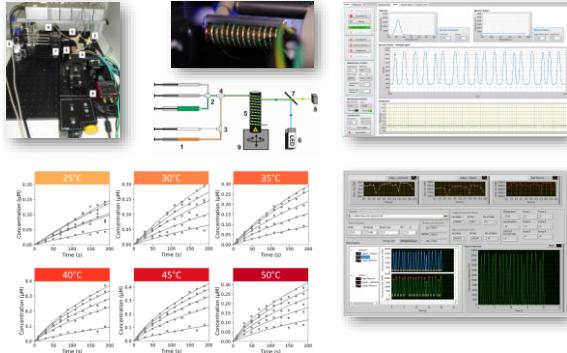


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Activity and specificity in μ-droplets



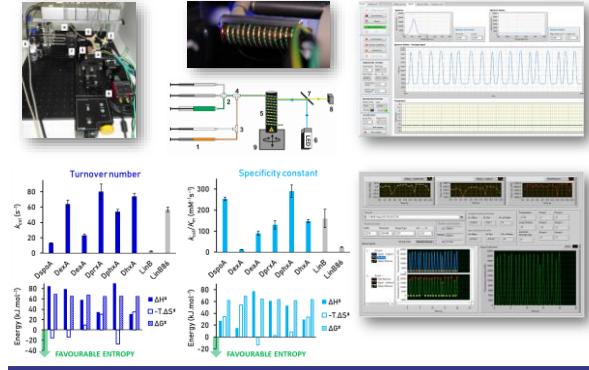
Kinetics and thermodynamics in μ -droplets



Vasina et al., *Chem Catalysis* 2: 2704

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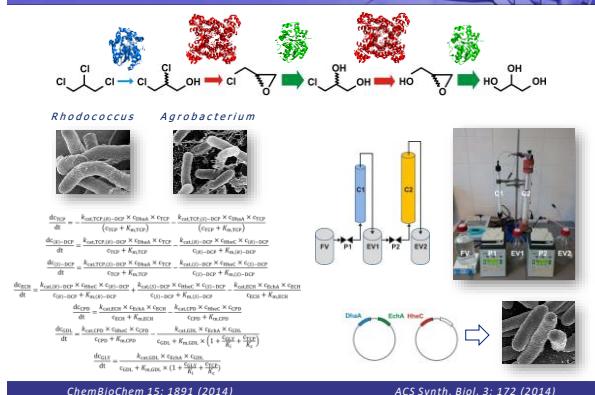
Kinetics and thermodynamics in μ -droplets



Vasina et al., *Chem Catalysis* 2: 2704

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Synthetic biology

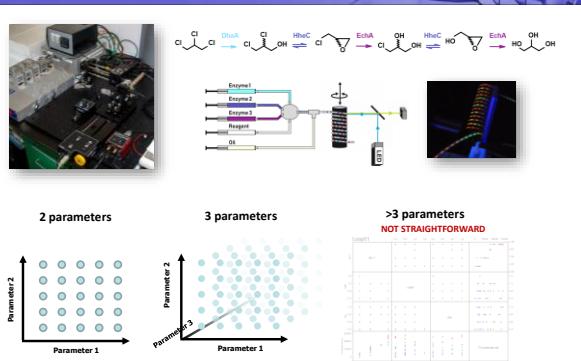


ChemBioChem 15: 1891 (2014)

ACS Synth. Biol. 3: 172 (2014)

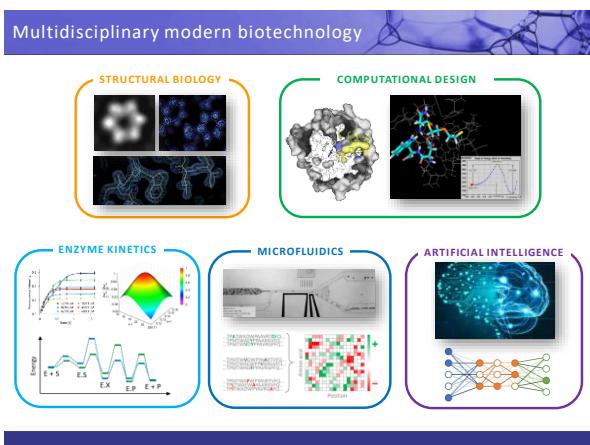
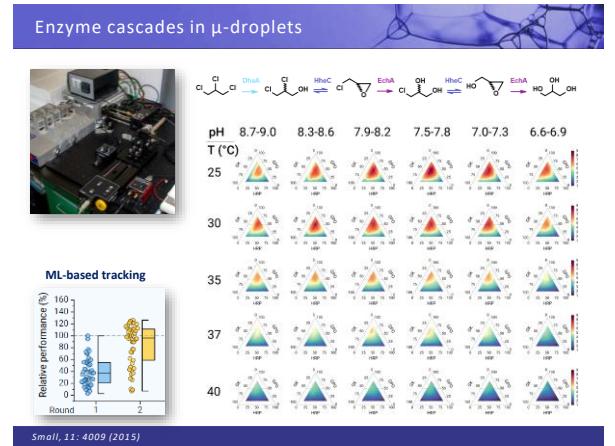
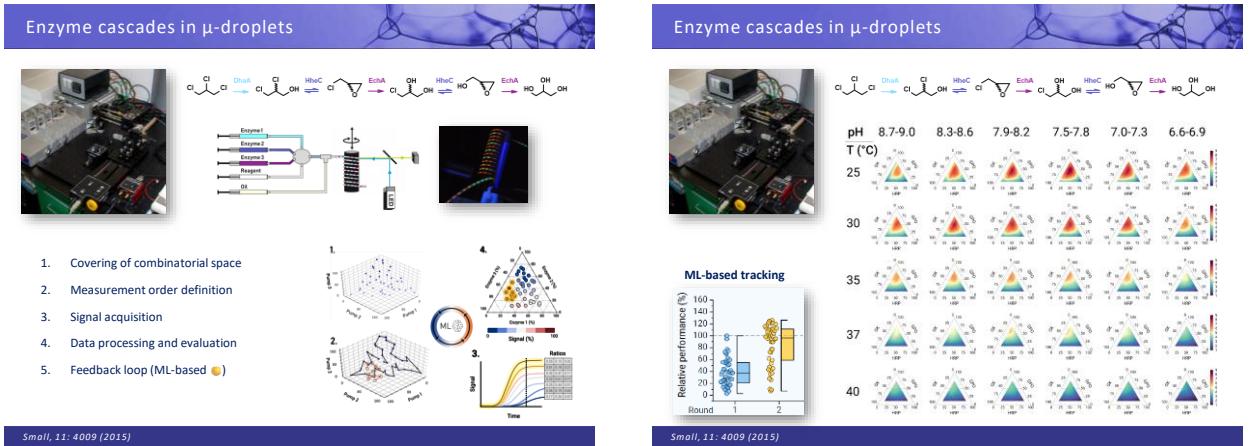
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Enzyme cascades in μ -droplets



Small, 11: 4009 (2015)

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Design and fabrication

□ direct fabrication methods

3D PRINTING



LASER CUTTING



CNC μ-MILLING



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Design and fabrication

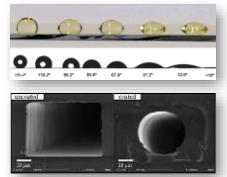
□ materials

- inert and transparent
- PDMS - poly(dimethyl siloxane)
- PMMA - poly(methyl methacrylate)
- fused silica, quartz and glass



□ surface modification

- plasma treatment
- silanization
- sol-gel coating



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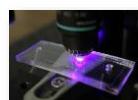
Sensing and detection

□ processing of small reagent volumes

□ analytical timescale and performance

□ on chip detection

- fluorescence (LSM, FCS, FLIM)
- UV/VIS absorbance
- IR spectroscopy
- Raman scattering
- (chemo/electro) luminescence
- thermal conductivity
- RI variation



□ off chip detection

- GC, HPLC, MS
- NMR, X-ray



Commercial instruments and services

□ customized design and fabrication



□ entire technologies



Nature Meth. 10, 1003 (2013)

Nature 499, 505 (2013)

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Conclusions

- ❑ reduced sample/reagent/power consumption
- ❑ superior performance and novel physics
- ❑ applications in life and medical sciences
- ❑ in-house as well as commercial technologies

microfluidics revolutionize science & technology

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Reading

- ❑ Mazurenko, S., 2020: **Machine Learning in Enzyme Engineering**. *ACS Catalysis*, 10, 1210–1223
- ❑ 3. DATABASES RELEVANT TO ENZYME ENGINEERING
- ❑ 3.3. Emerging Methods for High-Throughput Data Collection (page 1213 - 1216)



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pubs.acs.org/acscatalysis

Machine Learning in Enzyme Engineering

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