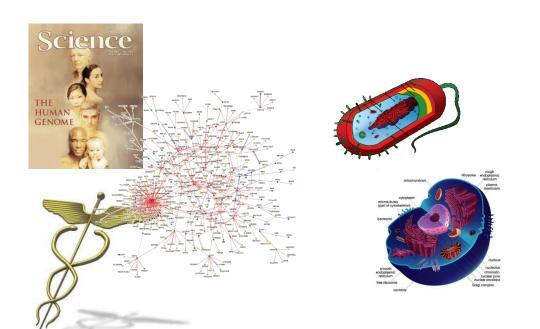
Proteomics & Protein-Protein Interactions

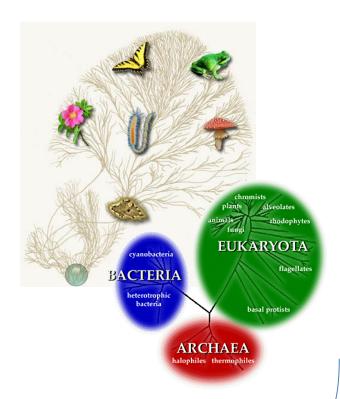
Jesse Rinehart, PhD
Biomedical Data Science: Mining & Modeling
CBB 752, Spring 2018





DNA -> RNA-> PROTEIN







DNA -> RNA-> PROTEIN

2007

Cel

Induction of Pluripotent Stem Cells from Adult Human Fibroblasts by Defined Factors

Kazutoshi Takahashi, Koji Tanabe, Mari Ohnuki, Megumi Narita, Comoko Ichisaka, Kiichiro Tomoda, and Shinya Yamanaka 1,2,3,4,*

2012

The Nobel Prize in Physiology or Medicine 2012 jointly to:

John B. Gurdon and Shinya Yamanaka "for the discovery that mature cells can be reprogrammed to become pluripotent"

DNA -> RNA-> PROTEIN

RNA-Guided Human Genome Engineering via Cas9

Prashant Mali, 1* Luhan Yang, 1,3* Kevin M. Esvelt, 2 John Aach, 1 Marc Guell, 1 James E. DiCarlo, 4 Julie E. Norville, 1 George M. Church 1,2+

Multiplex Genome Engineering Using CRISPR/Cas Systems

Le Cong, ^{1,2}* F. Ann Ran, ^{1,4}* David Cox, ^{1,3} Shuailiang Lin, ^{1,5} Robert Barretto, ⁶ Naomi Habib, ¹ Patrick D. Hsu, ^{1,4} Xuebing Wu, ⁷ Wenyan Jiang, ⁸ Luciano A. Marraffini, ⁸ Feng Zhang ¹†

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U.S. Summit Draws Attention to Technology with Potential, Peril

By Karen Pallarito (HealthDay News) Uploaded on December 21, 2015

Dec 2015

NATURE | NEWS

22 April 2015



Chinese scientists genetically modify human embryos

Rumours of germline modification prove true — and look set to reignite an ethical debate.

David Cyranoski & Sara Reardon

Hong Ma¹*, Nuria Marti-Gutierrez¹ Tomonari Hayama¹, Riffat Ahmed¹, Sang-Tae Kim2, Jianhui Gong5,6,3 Don P. Wolf1, Stephen B. Heitner10, J Shoukhrat Mitalipov1,108

Aug. 2017

doi:10.1038/nature23305

Correction of a pathogenic gene mutation in human embryos

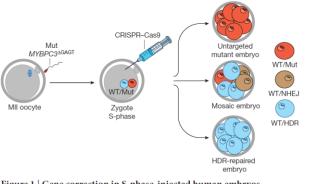
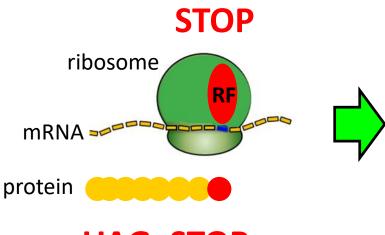


Figure 1 | Gene correction in S-phase-injected human embryos.

SYNTHETIC BIOLOGY DNA -> RNA-> PROTEIN

E. coli genome editing technologies to change321 native UAG stop codons to UAA and produced the First Whole Genome Edited Organism

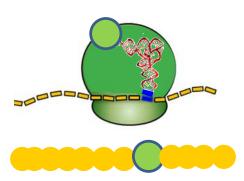


UAG=STOP

Native Genome

Uses UAG, UAA, UGA RF1= STOP at UAG





UAG=Sense

Native Genome

Uses only UAA, UGA and RF2 RF1 deletion

Can install new amino acid





Genomically Recoded Organisms Expand Biological Functions

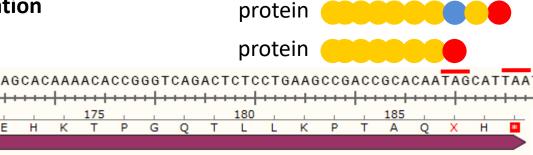
Marc J. Lajoie, ^{1,2} Alexis J. Rovner, ^{3,4} Daniel B. Goodman, ^{1,5} Hans-Rudolf Aerni, ^{4,6} Adrian D. Haimovich, ^{3,4} Gleb Kuznetsov, ¹ Jaron A. Mercer, ⁷ Harris H. Wang, ⁸ Peter A. Carr, ⁹ Joshua A. Mosberg, ^{1,2} Nadin Rohland, ¹ Peter G. Schultz, ¹⁰ Joseph M. Jacobson, ^{11,12} Jesse Rinehart, ^{4,6} George M. Church, ^{1,13*} Farren J. Isaacs^{3,4*}

SCIENCE VOL 342 18 OCTOBER 2013

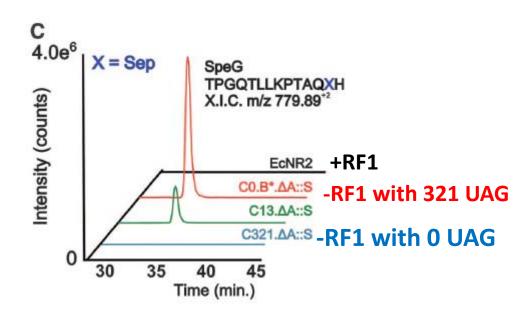
(Lajoie et al. Science 2013:PMID: 24136966)

Whole genome editing = Whole *proteome editing*

STOP at native UAG or translation to next in-frame TAA



Translation through 321 native UAG STOP codons was ablated with genome editing



(Lajoie et al. Science 2013:PMID: 24136966)

Proteomics

The study of the expression, location, modification, interaction, function, and structure of all the proteins in a given cell, organelle, tissue, organ, or whole organism.

Proteomics & Protein-Protein Interactions

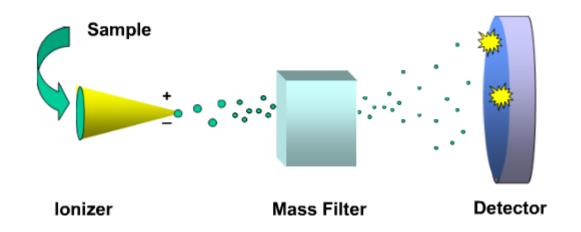
Overview

- Techniques & Technologies
 - Mass Spectrometry
 - Protein-Protein Interactions
 - Quantitative Proteomics
- Applications
 - Representative Studies
- Putting it all together....
 - Databases & Pathways

Principles of Mass Spectrometry (MS)

- In a mass spectrum we measure m/z (mass-to-charge)
- For proteins we measure peptide m/z
- A sample has to be ionizable in order to be analyzed

Basic Components of a Mass Spectrometer

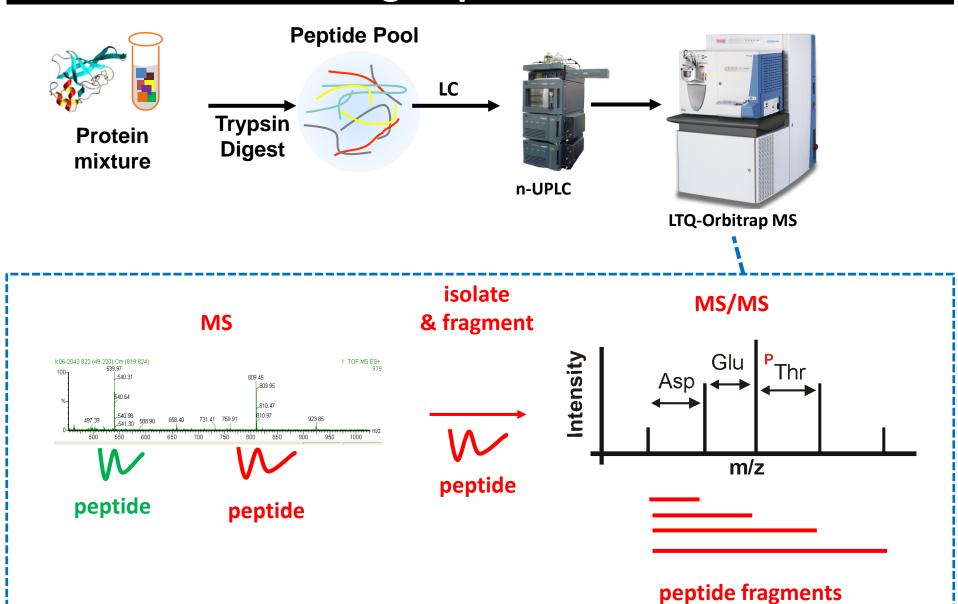




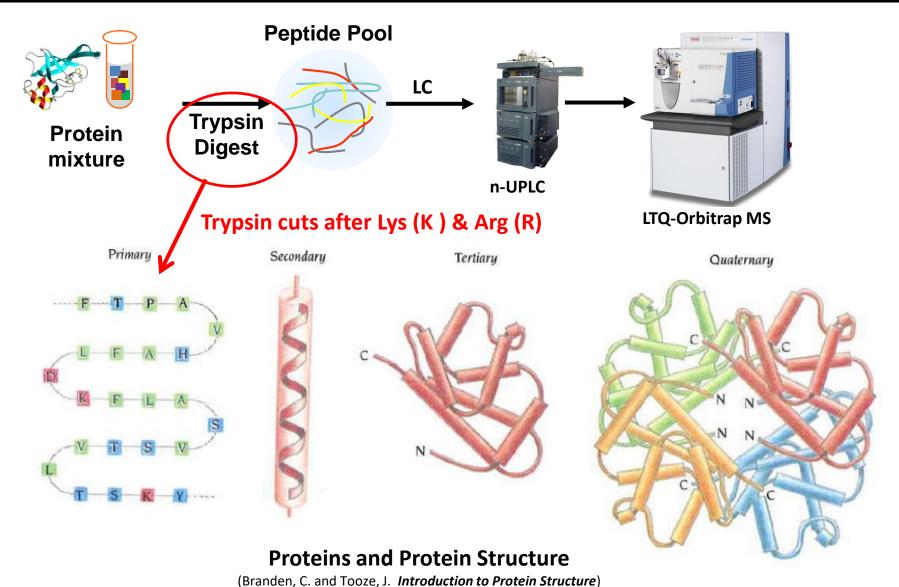
Two major <u>ionization</u> techniques enabled the success of mass spectrometry in the life sciences.

- Electrospray Ionization (ESI)
 Fenn JB, *Mann M, Meng CK, Wong SF, Whitehouse CM. Science. 1989
- Matrix Assisted Laser Desorption Ionization (MALDI)
 Tanaka K, Waki H, Ido Y, et al. Rapid Commun Mass Spectrom 1988
- 2002 Nobel Prize in Chemistry awarded to John B. Fenn & Koichi Tanaka
- Enabled direct measurement and "sequencing" of intact peptides & MS based Proteomics is born

Typical work flow for LC-MS "shotgun proteomics"



Typical work flow for LC-MS "shotgun proteomics"

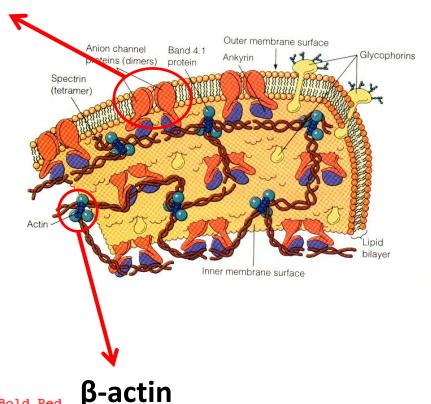


Trypsin digest followed by LC-MS: Examples of "Sequence Coverage"

Matched peptides shown in Bold Red

Band 3 Anion Transporter

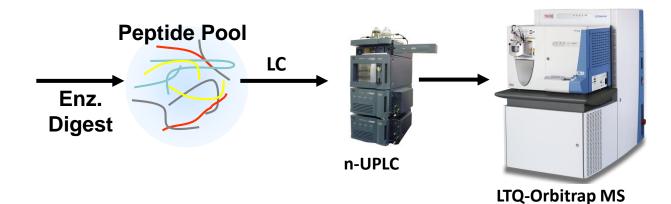
```
1 MEELQDDYED MMEENLEQEE YEDPDIPESQ MEEPAAHDTE ATATDYHTTS
 51 HPGTHKVYVE LQELVMDEKN QELRWMEAAR WVQLEENLGE NGAWGRPHLS
101 HLTFWSLLEL RRVFTKGTVL LDLQETSLAG VANQLLDRFI FEDQIRPQDR
151 EELLRALLLK HSHAGELEAL GGVKPAVLTR SGDPSQPLLP QHSSLETQLF
201 CEQGDGGTEG HSPSGILEKI PPDSEATLVL VGRADFLEOP VLGFVRLQEA
251 AELEAVELPV PIRFLFVLLG PEAPHIDYTQ LGRAAATLMS ERVFRIDAYM
301 AQSRGELLHS LEGFLDCSLV LPPTDAPSEQ ALLSLVPVQR ELLRRRYQSS
351 PAKPDSSFYK GLDLNGGPDD PLQQTGQLFG GLVRDIRRRY PYYLSDITDA
401 FSPOVLAAVI FIYFAALSPA ITFGGLLGEK TRNOMGVSEL LISTAVQGIL
451 FALLGAOPLL VVGFSGPLLV FEEAFFSFCE TNGLEYIVGR VWIGFWLILL
501 VVLVVAFEGS FLVRFISRYT QEIFSFLISL IFIYETFSKL IKIFODHPLQ
551 KTYNYNVLMV PKPQGPLPNT ALLSLVLMAG TFFFAMMLRK FKNSSYFPGK
601 LRRVIGDFGV PISILIMVLV DFFIQDTYTQ KLSVPDGFKV SNSSARGWVI
651 HPLGLRSEFP IWMMFASALP ALLVFILIFL ESQITTLIVS KPERKMVKGS
701 GFHLDLLLVV GMGGVAALFG MPWLSATTVR SVTHANALTV MGKASTPGAA
751 AQIOEVKEOR ISGLLVAVLV GLSILMEPIL SRIPLAVLFG IFLYMGVTSL
801 SGIQLFDRIL LLFKPPKYHP DVPYVKRVKT WRMHLFTGIQ IICLAVLWVV
851 KSTPASLALP FVLILTVPLR RVLLPLIFRN VELQCLDADD AKATFDEEEG
901 RDEYDEVAMP V
```



Matched peptides shown in Bold Red

```
1 MDDDIAALVV DNGSGMCKAG FAGDDAPRAV FPSIVGRPRH QGVMVGMGQK
51 DSYVGDEAQS KRGILTLKYP IEHGIVTNWD DMEKIWHHTF YNELRVAPEE
101 HPVLLTEAPL NPKANREKMT QIMFETFNTP AMYVAIQAVL SLYASGRTTG
151 IVMDSGDGVT HTVPIYEGYA LPHAILRLDL AGRDLTDYLM KILTERGYSF
201 TTTAEREIVR DIKEKLCYVA LDFEQEMATA ASSSSLEKSY ELPDGQVITI
251 GNERFRCPEA LFQPSFLGME SCGIHETTFN SIMKCDVDIR KDLYANTVLS
301 GGTTMYPGIA DRMQKEITAL APSTMKIKII APPERKYSVW IGGSILASLS
351 TFOOMWISKO EYDESGPSIV HRKCF
```





Mass Spectrum

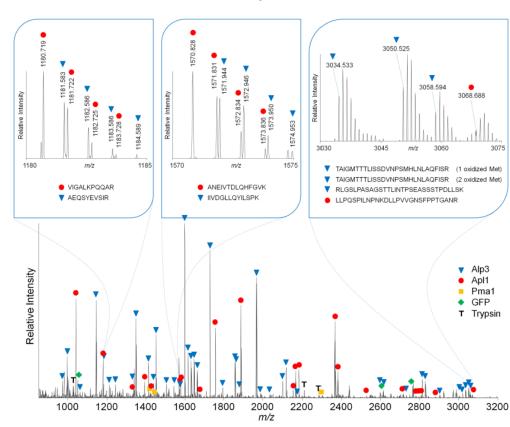
Peptide ions have a mass (m) and a charge (z).

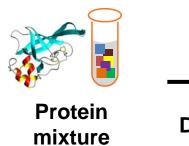
100 Da peptide:

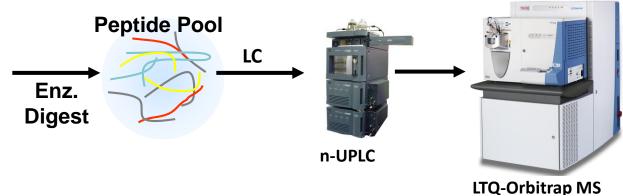
+1 = 100 m/z

+2 = 50 m/z

+3 = 33.3 m/z



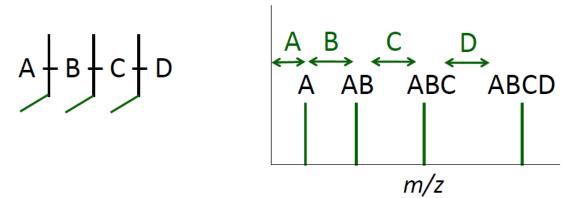


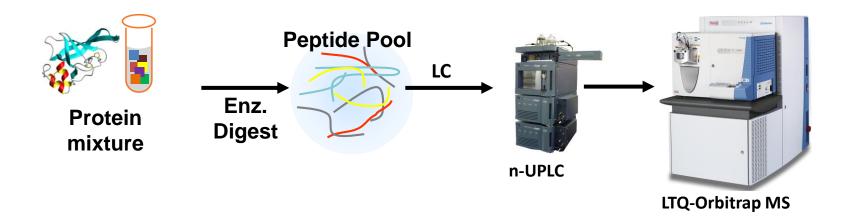


Peptide sequencing

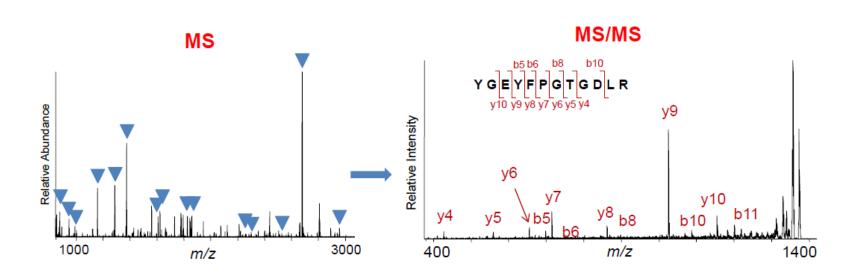
Peptide ions are isolated and "sequenced"

Simplified concept of peptide fragmentation





Database searching - at MS or MS/MS level



Computational Steps: massive amounts of MS data are read & interpreted. Databases searched to match peptide sequences.

The *pace of proteomics is set by a combination of techniques and technological advances.

*orders of magnitude behind genome technologies (sequencing)

Yeast proteome reported in Washburn et al. Nature Biotech 2001:

~82 hours* = 1,484 proteins - ~0.3 proteins/ min

*estimates from paper: 3 fractions @ 15 X 110 minute "runs" for each fraction

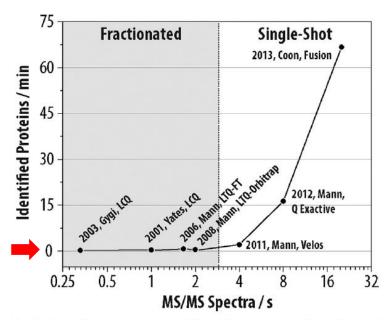


Fig. 5. Rate of protein identifications as a function of mass spectrometer scan rate for selected large-scale yeast proteome analyses over the past decade. Each data point is annotated with the year, corresponding author, type of MS system used, and reference number.

The *pace of proteomics is set by a combination of techniques and technological advances.

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Technological Innovation and Resources

* Author's Choice

© 2014 by The American Society for Biochemistry and Molecular Biology, Inc. This paper is available on line at http://www.mcponline.org

Alexander S. Hebert‡§**, Alicia L. Richards§¶**, Derek J. Bailey§¶, Arne Ulbrich§¶, Emma E. Coughlin§, Michael S. Westphall§, and Joshua J. Coon‡§¶∥

On average, each **one hour** analysis achieved detection of **3,977 proteins**

PROTOCOL

One-hour proteome analysis in yeast

Alicia L Richards^{1,2,4}, Alexander S Hebert^{1,3,4}, Arne Ulbrich^{1,2}, Derek J Bailey^{1,2}, Emma E Coughlin¹, Michael S Westphall¹ & Joshua J Coon^{1–3}

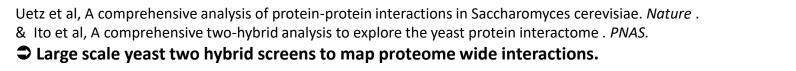
"...the identification of up to **4,002 proteins**, This protocol, which includes cell lysis, overnight tryptic digestion, sample analysis and database searching, **takes ~24 h to complete**."

The one hour yeast proteome. Hebert AS, et a, Coon JJ.

Mol Cell Proteomics. 2014 PMID: 24143002 & Nat Protoc. 2015. PMID: 25855955

A tour of proteomics: Studies with the budding yeast Saccharomyces cerevisiae

2000 & 2001



2001

Washburn, et al. Large-scale analysis of the yeast proteome by multidimensional protein identification technology. *Nature Biotechnol*.

⇒ Established the 'shotgun' technology by showing that many proteins in a yeast-cell lysate could be identified in a single experiment.

2002

Ho, Y. et al. Systematic identification of protein complexes in Saccharomyces cerevisiae by mass spectrometry. Nature.

- & Gavin, A. C. et al. Functional organization of the yeast proteome by systematic analysis of protein complexes. Nature.
- **⇒** Protein–protein interaction maps can be obtained by MS; the yeast cell is organized into protein complexes.

2003

Ghaemmaghami, S. et al. Global analysis of protein expression in yeast. *Nature*. **&** Huh, W. K. et al. Global analysis of protein localization in budding yeast. *Nature*.

⇒ TAP-Tag and expression studies & GFP-Tag and localization studies

2006

Krogan NJ, et al. Global landscape of protein complexes in the yeast Saccharomyces cerevisiae. Nature.

⇒ TAP-Tag and Protein-Protein Interaction

2008

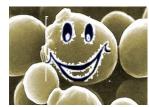
de Godoy LM, et al. Comprehensive mass-spectrometry-based proteome quantification of haploid versus diploid yeast. *Nature*.

⇒ SILAC based quantitation of an entire proteome.

2009

Picotti P, et al. Full dynamic range proteome analysis of S. cerevisiae by targeted proteomics. Cell.

Towards proteome wide targeted proteomics.



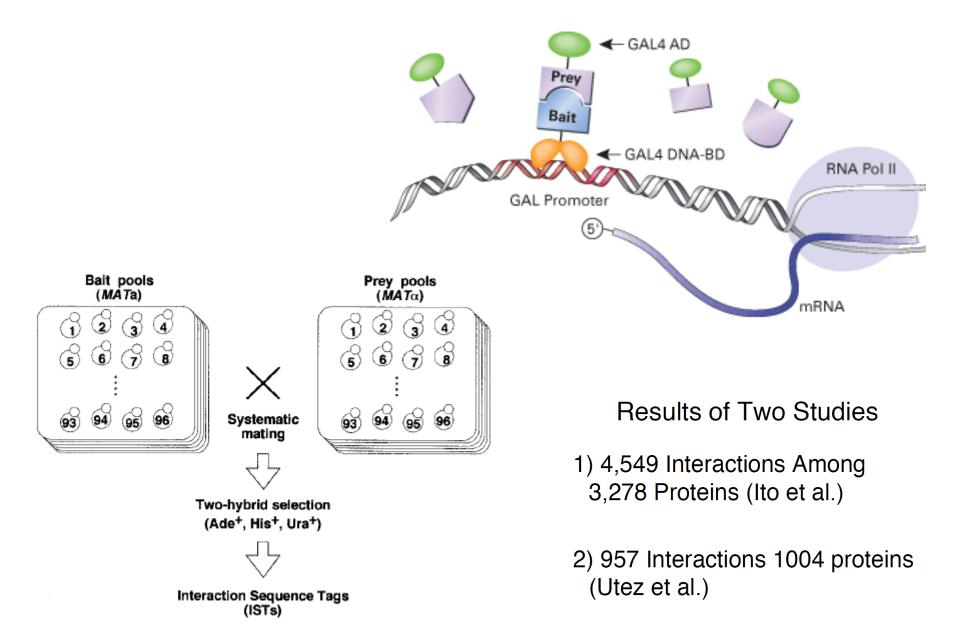
A comprehensive analysis of protein-protein interactions in Saccharomyces cerevisiae.

Uetz et al, Nature 2000 **Yeast Two Hybrid Assay** Ito et al, PNAS 2001 ← GAL4 AD Prev Bait ◆─ GAL4 DNA-BD RNA Pol II **GAL Promoter** Clone bait and prey constructs and place in separate strains. Mat a Mat α mRNA Gal4 DNA Protein B Protein A **Activation Domain Binding Domain Fusion Proteins** Gal4 DNA HIS3 Selectable Marker **Binding Sites** Mate a + α If A & B interact **Colonies Grow** On Plates Lacking

Histidine

HIS3 Expression

Uetz et al, Nature 2000 Ito et al, PNAS 2001



A comprehensive analysis of protein-protein interactions in Saccharomyces cerevisiae.

Uetz et al, Nature 2000 Ito et al, PNAS 2001

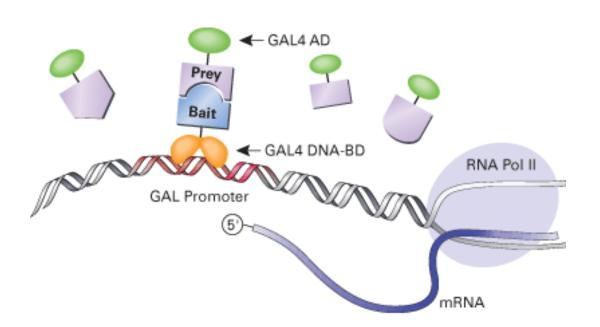
Yeast Two Hybrid Assay

Advantages:

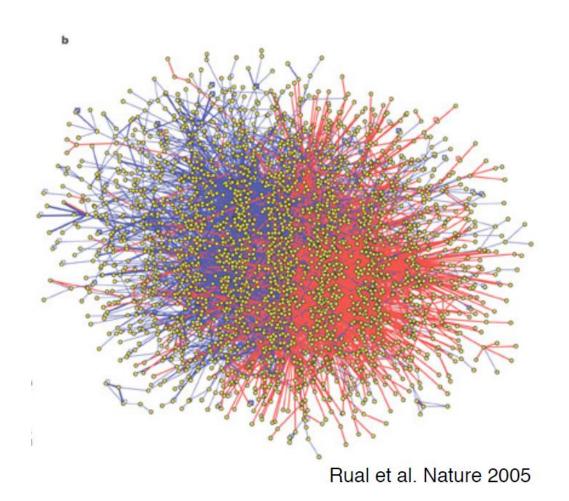
- In vivo assay
- Simple

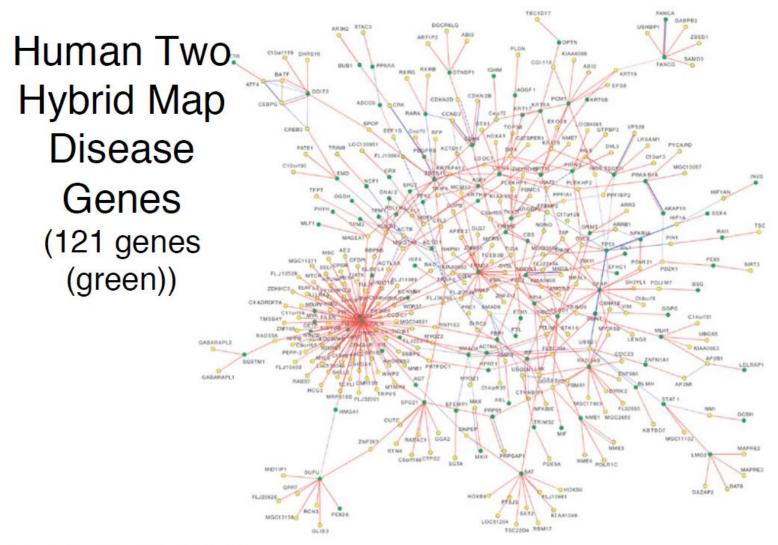
Some Disadvantages

- Hard to execute on large scale
- False positives: a real interaction or "possible" interaction
- Interaction in nucleus (required for GAL system)
- Clones are fusion proteins and sometimes "partial" proteins
- Multiple protein complexes not "captured"



Human Two Hybrid Map 8,100 ORFs (~7,200 genes) 10,597 interactions

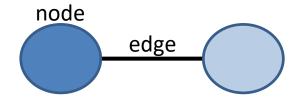


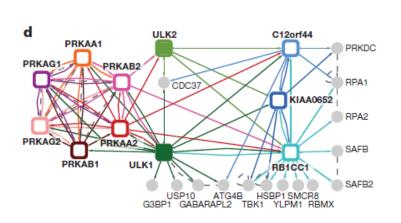


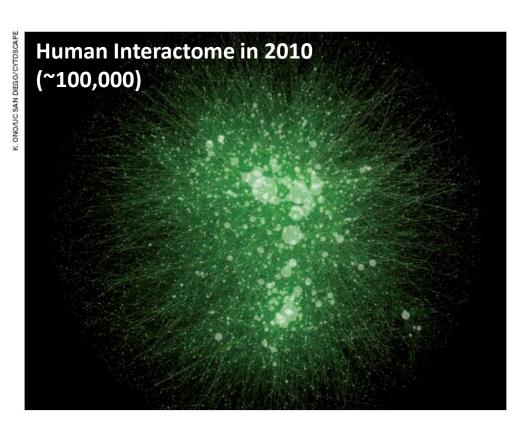
Rual et al. Nature 2005 Vol 437

Protein-Protein interaction maps:

Proteins are represented by **nodes** and interactions are represented by **edges** between nodes.

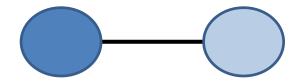






Bonetta, Nature 2010

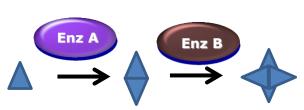
Protein-Protein interactions:

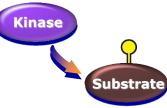


Some examples:

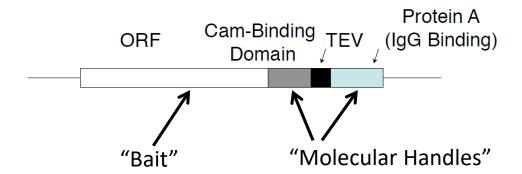
- Physical and direct
- Physical and indirect
 - Multi-protein complexes
 - Scaffolds
- Transient
 - Kinase & substrate

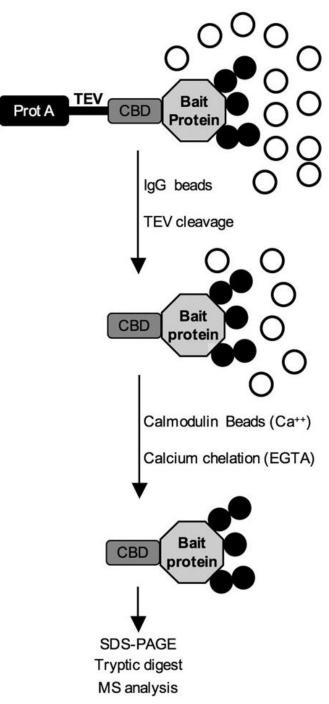




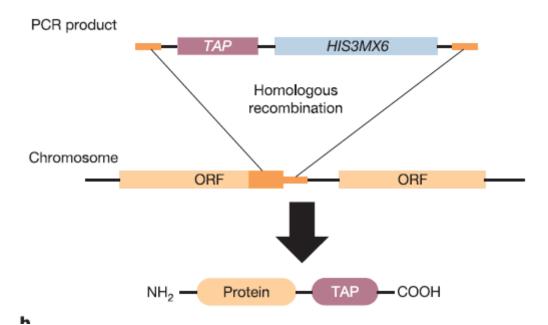


Tandem Affinity Purification (TAP) Tagging





Global TAP Tagging in yeast



2003

Ghaemmaghami, S. et al. Global analysis of protein expression in yeast. *Nature*. **&** Huh, W. K. et al. Global analysis of protein localization in budding yeast. *Nature*.

⇒ TAP-Tag and expression studies & GFP-Tag and localization studies

2002

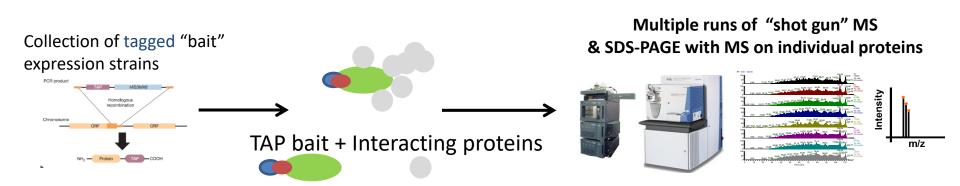
Ho, Y. et al. Systematic identification of protein complexes in Saccharomyces cerevisiae by mass spectrometry. Nature.

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- ⇒ Protein—protein interaction maps can be obtained by MS; the yeast cell is organized into protein complexes.

2006

Krogan NJ, et al. Global landscape of protein complexes in the yeast Saccharomyces cerevisiae. Nature.

⇒ TAP-Tag and Protein-Protein Interaction



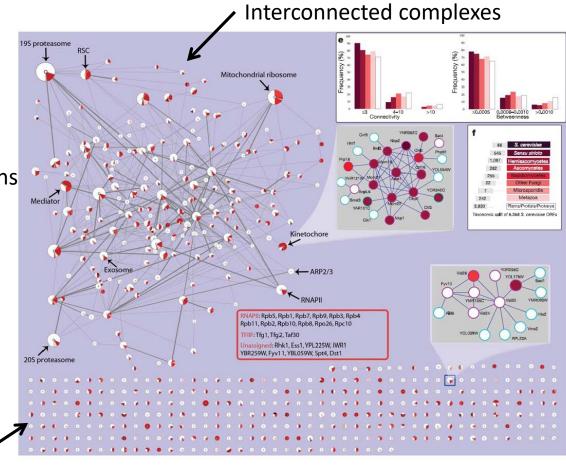
Krogan et al. observed 7,123 protein-protein interactions:

Important aspects:

- Tagged the native genes and did not overexpress the fusion proteins
- Could immediately validate partners (reciprocal purification in data set)
- Complementary MS techniques, deeper coverage of complexes
- Authors state, "...rigorous computational procedures to assign confidence values to our predictions..."

Cellular proteins are organized into complexes

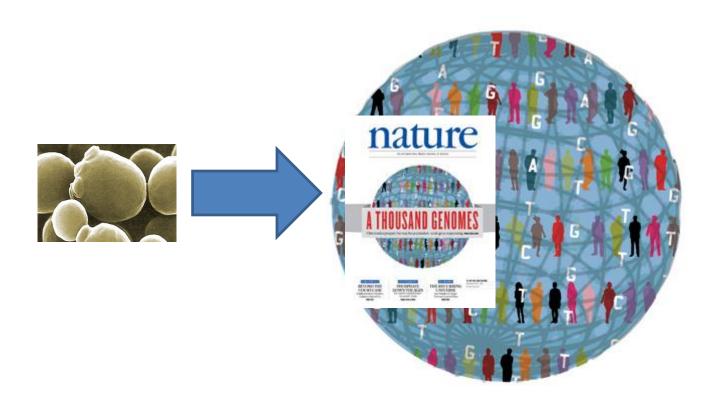
- 4,562 tagged proteins
- 2,357 successful purifications
- Identified 4,087 interacting proteins
 ~72 % proteome
- Majority of the yeast proteome is organized into complexes
- Many complexes are conserved in other species



Krogan NJ, et al. Nature. 2006

Complexes with little or no interconnectivity

How do we learn more about the organization of the human proteome?

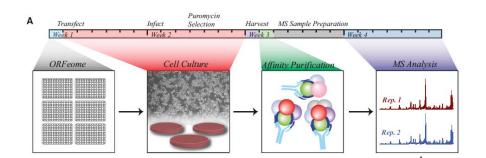


BioPlex (Biophysical Interactions of ORFeome-derived complexes)

~25% of human genes used as baits

5,891 IP-MS experiments

56,553 interactions from 10,961 proteins



The BioPlex Network: A Systematic Exploration of the Human Interactome

Edward L. Huttlin,¹ Lily Ting,¹ Raphael J. Bruckner,¹ Fana Gebreab,¹ Melanie P. Gygi,¹ John Szpyt,¹ Stanley Tam,¹ Gabriela Zarraga,¹ Greg Colby,¹ Kurt Baltier,¹ Rui Dong,² Virginia Guarani,¹ Laura Pontano Vaites,¹ Alban Ordureau,¹ Ramin Rad,¹ Brian K. Erickson,¹ Martin Wühr,¹ Joel Chick,¹ Bo Zhai,¹ Deepak Kolippakkam,¹ Julian Mintseris,¹ Robert A. Obar,¹.³ Tim Harris,³ Spyros Artavanis-Tsakonas,¹.³ Mathew E. Sowa,¹ Pietro De Camilli,² Joao A. Paulo,¹ J. Wade Harper,¹.* and Steven P. Gygi¹.*

BioPlex 1.0 Huttlin et al, Cell. 2015, PMID: 26186194

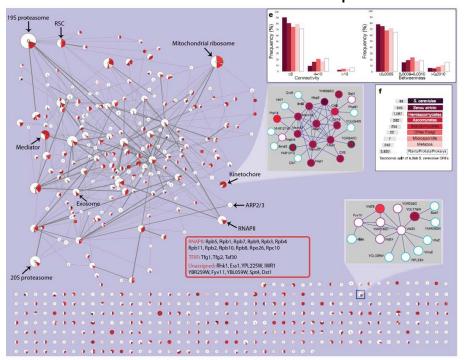
Architecture of the human interactome defines protein communities and disease networks

Edward L. Huttlin¹, Raphael J. Bruckner¹, Joao A. Paulo¹, Joe R. Cannon¹, Lily Ting¹, Kurt Baltier¹, Greg Colby¹, Fana Gebreab¹, Melanie P. Gygi¹, Hannah Parzen¹, John Szpyt¹, Stanley Tam¹, Gabriela Zarraga³, Laura Pontano-Vaires¹, Sharan Swarup¹, Anne E. White¹, Devin K. Schweppe¹, Ramin Rad¹, Brian K. Erickson¹, Robert A. Obar^{1,2}, K. G. Guruharsha², Kejie Li², Spyros Artavanis -Tsakonas^{1,2}, Steven P. Gygi¹ & J. Wade Harper¹

BioPlex 2.0 Huttlin et al, Nature. 2017 PMID: 28514442

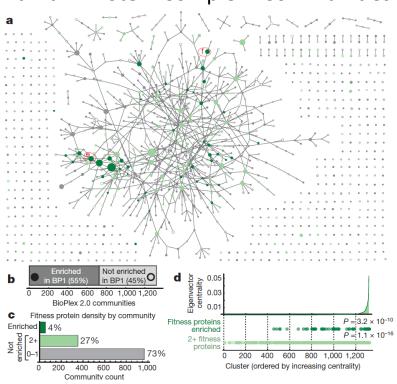
Cellular proteins are organized into complexes and this proteome organization is conserved

Yeast: Interaction Network of Complexes



Krogan NJ, et al. *Nature*. 2006 PMID: 16554755

Human: Protein Complex "Communities"

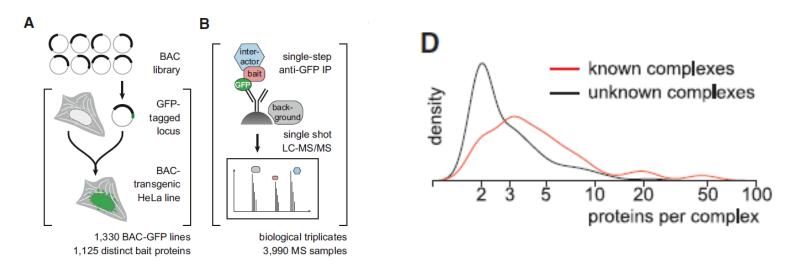


Huttlin et al, Nature. 2017 PMID: 28514442

A Human Interactome in Three Quantitative Dimensions Organized by Stoichiometries and Abundances

Marco Y. Hein, 1,6,8 Nina C. Hubner, 1,6,9 Ina Poser, 2 Jürgen Cox, 1 Nagarjuna Nagaraj, 1 Yusuke Toyoda, 2,10 Igor A. Gak, 3 Ina Weisswange, 4,5 Jörg Mansfeld, 3 Frank Buchholz, 2,4 Anthony A. Hyman, 2,7,* and Matthias Mann 1,7,*

- GFP-tagged proteins are expressed in mammalian cell lines from BAC transgenes with near-endogenous expression patterns
- Human interactome dataset connecting **5,400** proteins with **28,500** interactions



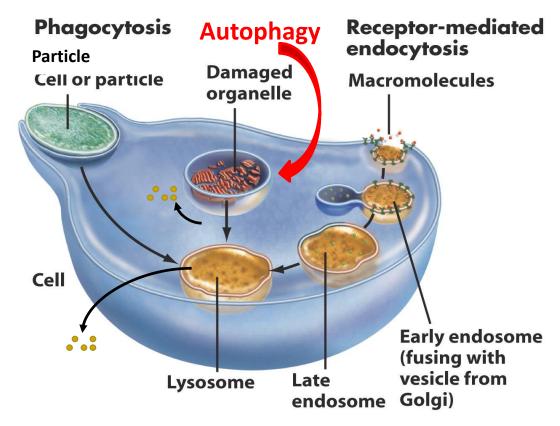
- Three quantitative dimensions measure specificities, stoichiometries, and abundances
- Stable complexes are rare but stand out by a signature of balanced stoichiometries
- Weak interactions dominate the network and have critical topological properties

Hein MY, et al. Mann M, Cell. 2015 PMID: 26496610

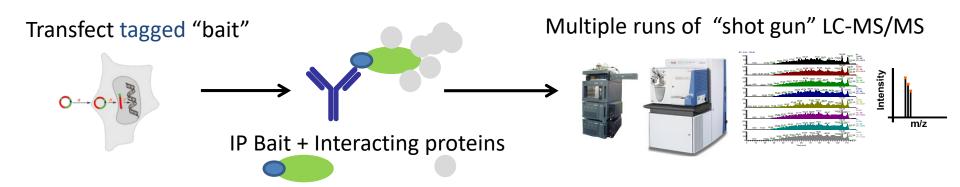
ARTICLES

Network organization of the human autophagy system

Christian Behrends¹, Mathew E. Sowa¹, Steven P. Gygi² & J. Wade Harper¹



Pearson Prentice Hall, Inc. 2005 www.stolaf.edu/people/giannini/cell/lys.htm



~65 bait proteins LC-MS/MS identifies 2553 proteins

Data analysis to sort out real interaction from background

Authors use CompPASS to identify High-Confidence Interacting Proteins (HCIP)

763 HCIPs identified that compose The Autophagy Interaction Network

Autophagy Interaction Network

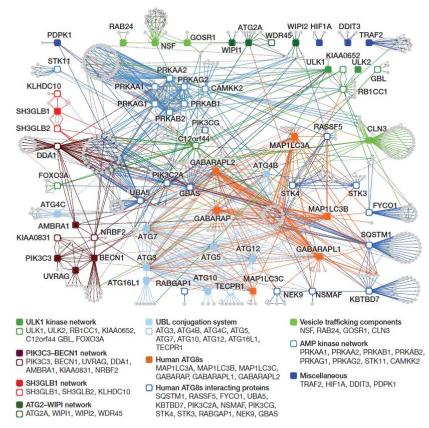


Figure 1 Overview of the autophagy interaction network (AIN). HCIPs within the autophagy network are shown for 32 primary baits (filled squares) and 33 secondary baits (open squares). Subnetworks are colour-coded. Interacting proteins are indicated by grey circles.

The Hippo Signaling Pathway Interactome

Young Kwon,¹ Arunachalam Vinayagam,¹* Xiaoyun Sun,³* Noah Dephoure,⁴ Steven P. Gygi,⁴ Pengyu Hong,³ Norbert Perrimon^{1,2}†

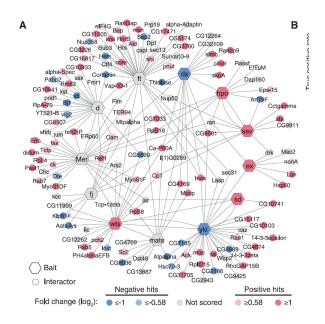
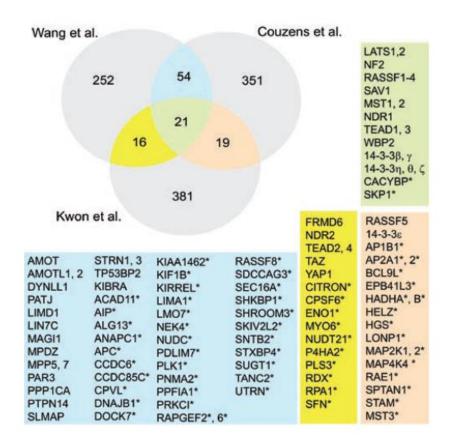


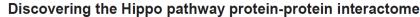
Fig. 2. Validation of Hippo-PPIN with functional RNAi screen and co-IP. (**A**) Distribution of Yki-reporter values for individual double-stranded RNAs (dsRNAs) in our focused RNAi screen. About 70% of genes are covered by two dsRNAs. (**B**) Recovery of Hippo pathway components from RNAi screen [fold-change (\log_2) cutoff \pm 1]. (**C**) The positive



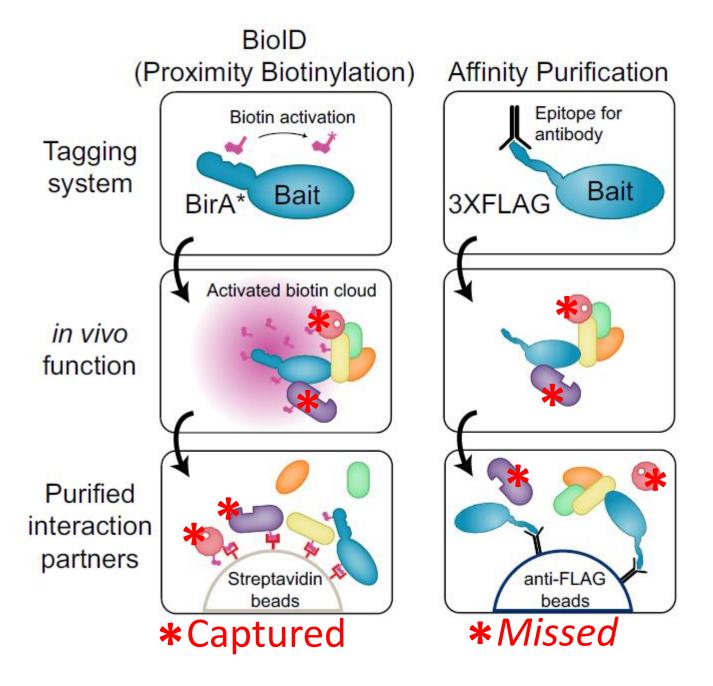
RESEARCH HIGHLIGHT

Cell Research (2014) 24:137-138.

© 2014 IBCB, SIBS, CAS All rights reserved 1001-0602/14 \$ 32.00



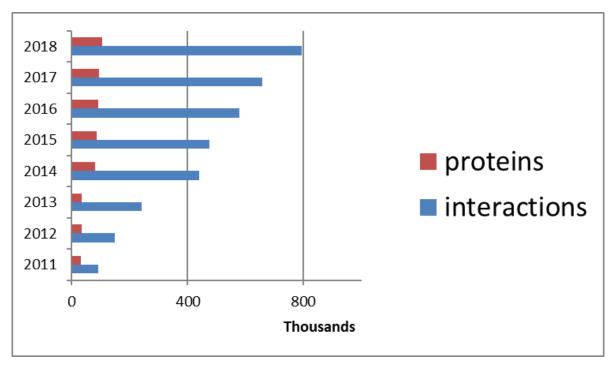
Cell Research (2014) 24:137-138. doi:10.1038/cr.2014.6; published online 14 January 2014

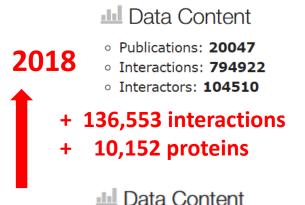


Protein-Protein Interaction Databases



http://www.ebi.ac.uk/intact/





o Publications: 14451

Interactions: 658369
 Interactors: 94358

2017

Proteomics & Protein-Protein Interactions

Overview

- Techniques & Technologies
 - Mass Spectrometry
 - Protein-Protein Interactions
 - Quantitative Proteomics
- Applications
 - Representative Studies
- Putting it all together....
 - Databases & Pathways

Protein interaction networks:

Some of the many important aspects:

- Parts List
- Organization and assembly
- Biological function can be inferred



However:

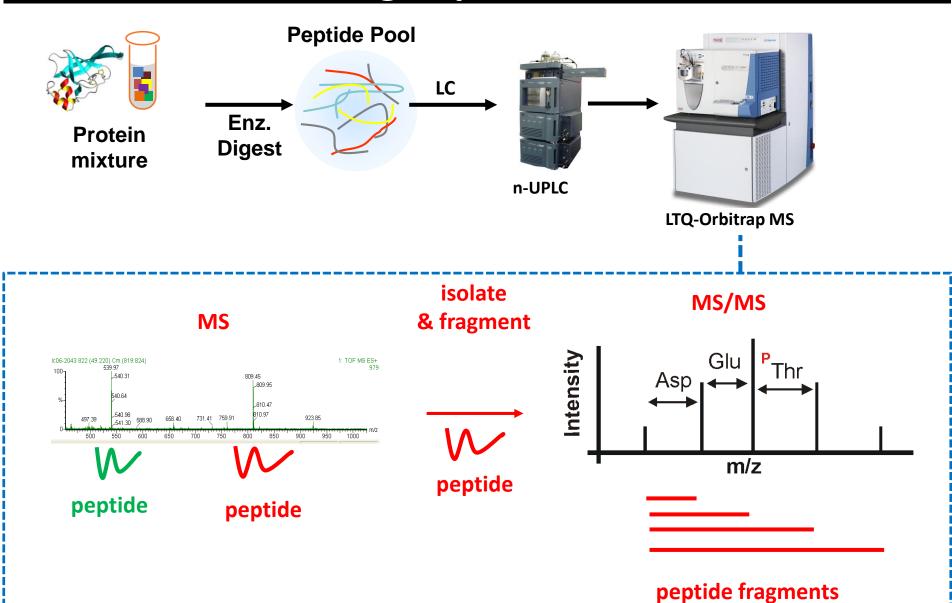
- Interaction data is largely static

Next Step:

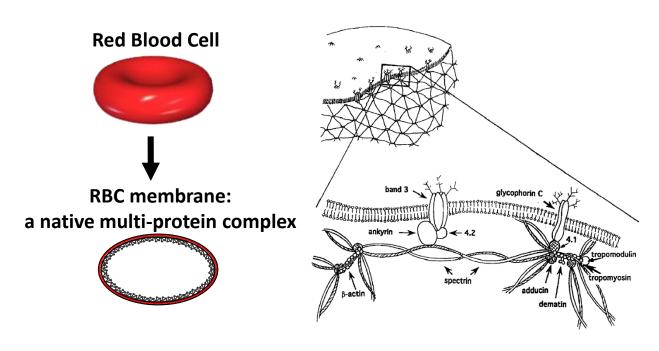
- How do protein interaction networks change over time?



Typical work flow for LC-MS "shotgun proteomics"

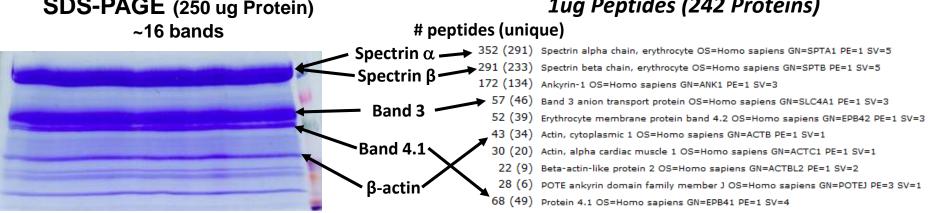


MS Data is <u>not</u> inherently quantitative, but ...



RBC membrane proteome Coomassie Stained SDS-PAGE (250 ug Protein)

RBC membrane proteome Shotgun Proteomics 1ug Peptides (242 Proteins)



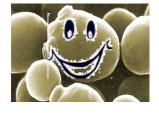
Rinehart et al., unpublished

A tour of proteomics: Studies with the budding yeast Saccharomyces cerevisiae

2000 & 2001

Uetz et al, A comprehensive analysis of protein-protein interactions in Saccharomyces cerevisiae. Nature.

- & Ito et al, A comprehensive two-hybrid analysis to explore the yeast protein interactome . PNAS.
- **□** Large scale yeast two hybrid screens to map proteome wide interactions.



2001

Washburn, et al. Large-scale analysis of the yeast proteome by multidimensional protein identification technology. *Nature Biotechnol*.

⇒ Established the 'shotgun' technology by showing that many proteins in a yeast-cell lysate could be identified in a single experiment.

2002

Ho, Y. et al. Systematic identification of protein complexes in Saccharomyces cerevisiae by mass spectrometry. Nature.

- & Gavin, A. C. et al. Functional organization of the yeast proteome by systematic analysis of protein complexes. Nature.
- **⇒** Protein–protein interaction maps can be obtained by MS; the yeast cell is organized into protein complexes.

2003

Ghaemmaghami, S. et al. Global analysis of protein expression in yeast. *Nature*. **&** Huh, W. K. et al. Global analysis of protein localization in budding yeast. *Nature*.

⇒ TAP-Tag and expression studies & GFP-Tag and localization studies

2006

Krogan NJ, et al. Global landscape of protein complexes in the yeast Saccharomyces cerevisiae. Nature.

⇒ TAP-Tag and Protein-Protein Interaction

2008

de Godoy LM, et al. Comprehensive mass-spectrometry-based proteome quantification of haploid versus diploid yeast. Nature.

⇒ SILAC based quantitation of an entire proteome.

2009

Picotti P, et al. Full dynamic range proteome analysis of S. cerevisiae by targeted proteomics. Cell.

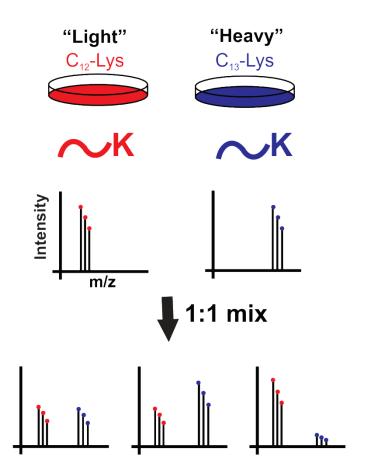
Towards proteome wide targeted proteomics.

Quantitative Proteomics

S.I.L.A.C. - Stable isotope labeling with amino acids in cell culture

-Ong S.E. et al. *Molecular & Cell Proteomics* 2002

- Stable isotopes are *not radioactive*, and they occur naturally in nature. For example, 99% of all carbon in the world is carbon-12 (¹²C) and 1% is carbon-13 (¹³C).
- SILAC reagents have enriched stable isotopes that have been placed into compounds in abundances much greater than their natural abundance.
- We can obtain labeled compounds with ~95-99% ¹³C.
- Because a mass spectrometer separates ions by mass, we use mass spectrometry to distinguish isotopes in compounds by their mass.
- Simultaneous comparison in the same MS run is <u>key</u>



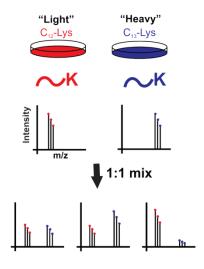
2008

de Godoy LM, et al. Comprehensive mass-spectrometry-based proteome quantification of haploid versus diploid yeast. Nature.

SILAC based quantitation of an entire proteome.

S.I.L.A.C. - Stable isotope labeling with amino acids in cell culture

-Ong SE et al. Molecular & Cell Proteomics 2002.

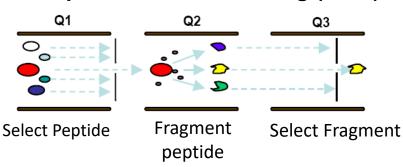


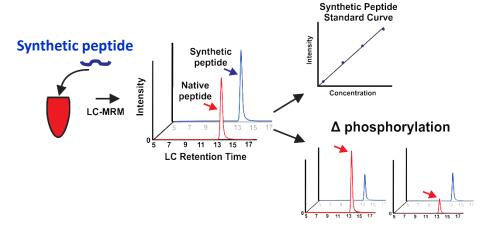
2009

Picotti P, et al. Full dynamic range proteome analysis of S. cerevisiae by targeted proteomics. Cell.

Towards proteome wide targeted proteomics.

Multiple Reaction Monitoring (MRM)





2008

de Godoy LM, et al. Comprehensive mass-spectrometry-based proteome quantification of haploid versus diploid yeast. *Nature*. 30;455(7217):1251-4.

SILAC based quantitation of an entire proteome.

Table 1 | Yeast ORFs identified by SILAC-based quantitative proteomics

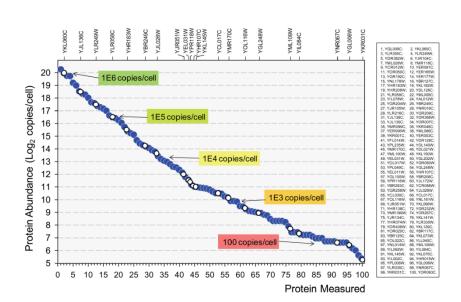
	Number of ORFs	TAP	GFP	nanoLC-MS
Total yeast ORFs Characterized yeast ORFs	6,608 4,666	4,251 3,629	4,154 3,581	4,399 3,824
Uncharacterized yeast ORFs	1,128	581	539	572
Dubious yeast ORFs Not present in ORF database	814	26 (3%) 15	23 (3%) 11	3 (<1%) 0

Comparative sequencing shows that 814 of the 6,608 yeast ORFs are never expressed (dubious ORFs, http://www.yeastgenome.org). Of these only six were identified in this experiment and three were validated by SILAC-assisted *de novo* sequencing of several peptides (Supplementary Table 5 and Supplementary Figs 2–4). Two of the three validated ones were reclassified as genuine yeast genes during writing of this manuscript (YGL041W-A and YPR170W-B). This leaves three potential false-positives (0.37% of 815) and suggests that our estimate of a false-positive identification rate of maximally 1% is conservative.

2009

Picotti P, et al. Full dynamic range proteome analysis of S. cerevisiae by targeted proteomics. Cell.

Towards proteome wide targeted proteomics.



2008

de Godoy LM, et al. Comprehensive mass-spectrometry-based proteome quantification of haploid versus diploid yeast. *Nature*. SILAC based quantitation of an entire proteome.

Pheromone signaling is required for mating of haploid cells and is absent from diploid cells.

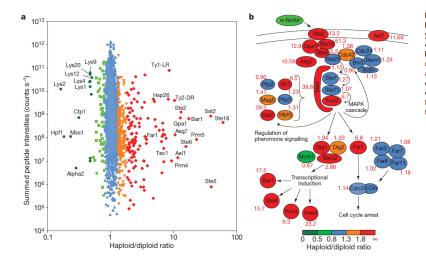


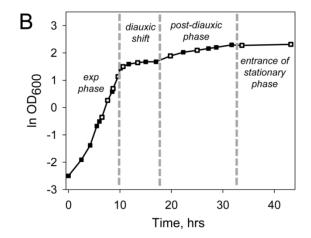
Figure 3 | Quantitative differences between the haploid and diploid yeast proteome. a, Overall fold change for the yeast proteome. B. Members of the yeast pheromone response are colour-coded according to fold change. The diploid to haploid ratio as determined by SILAC is indicated for each protein. Figure is adapted from ref. 13.

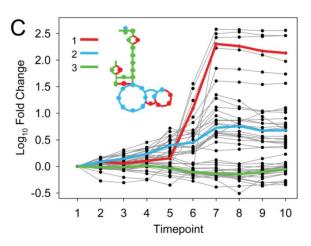
2009

Picotti P, et al. Full dynamic range proteome analysis of S. cerevisiae by targeted proteomics. Cell.

Towards proteome wide targeted proteomics.

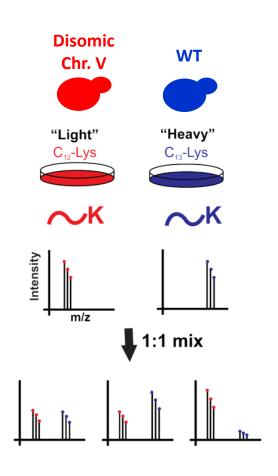
Network expression dynamics

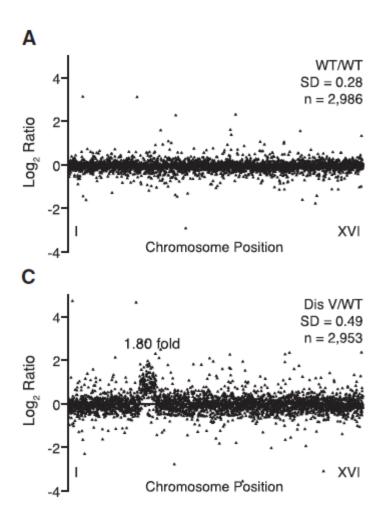




Identification of Aneuploidy-Tolerating Mutations

Eduardo M. Torres,^{1,2} Noah Dephoure,³ Amudha Panneerselvam,¹ Cheryl M. Tucker,⁴ Charles A. Whittaker,¹ Steven P. Gygi,³ Maitreya J. Dunham,⁵ and Angelika Amon^{1,2,*}



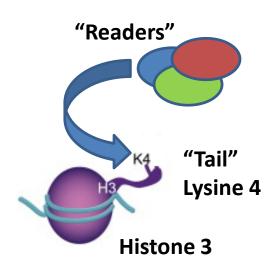


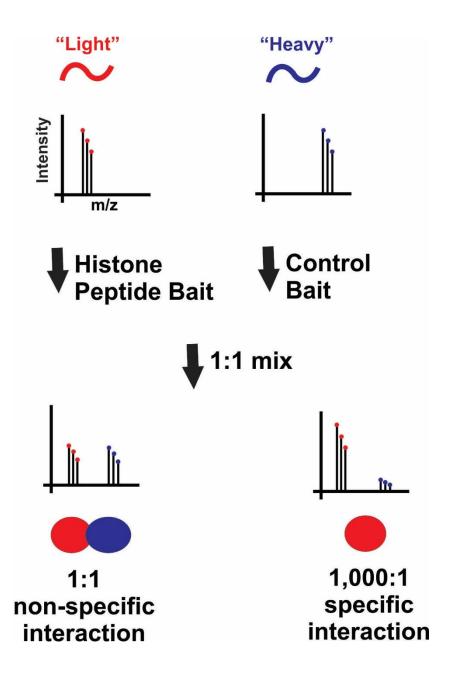


Resource

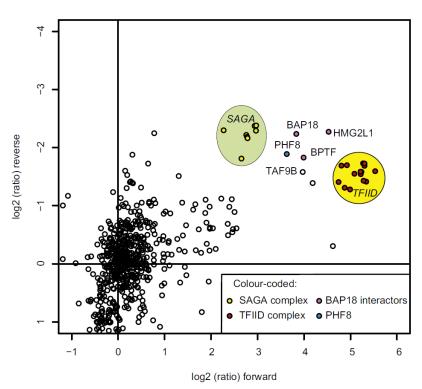
Quantitative Interaction Proteomics and Genome-wide Profiling of Epigenetic Histone Marks and Their Readers

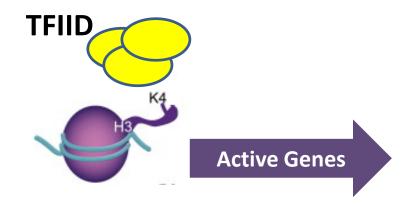
Michiel Vermeulen,^{1,6,7,*} H. Christian Eberl,^{1,6} Filomena Matarese,^{2,6} Hendrik Marks,² Sergei Denissov,² Falk Butter,¹ Kenneth K. Lee,³ Jesper V. Olsen,^{1,5} Anthony A. Hyman,⁴ Henk G. Stunnenberg,^{2,*} and Matthias Mann^{1,*}



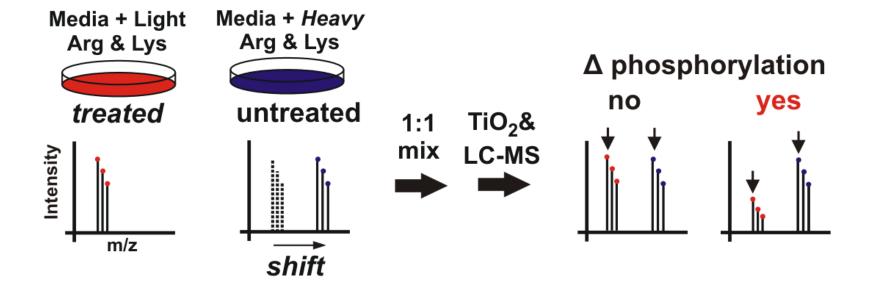


H3K4me3 interactors

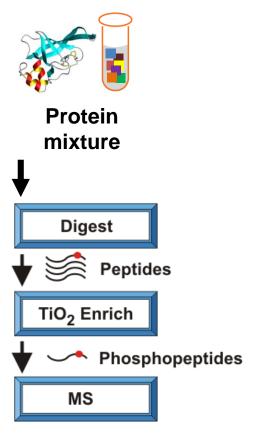




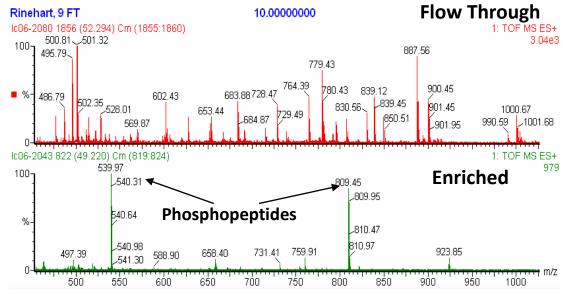
A SILAC approach to study protein phosphorylation dynamics



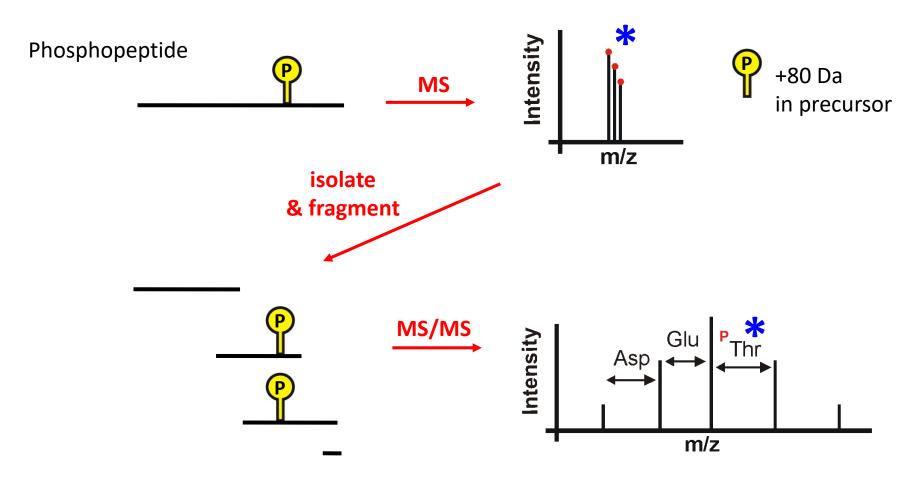
Major technological advances in mass spectrometers and phosphopeptide enrichment



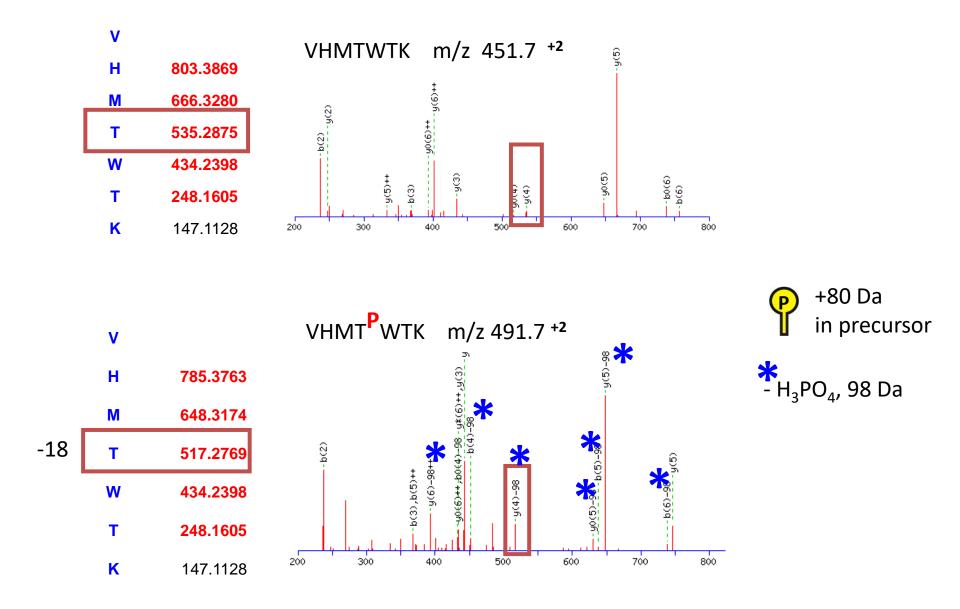
TiO₂ Enrichment



*Phosphopeptide signatures in MS



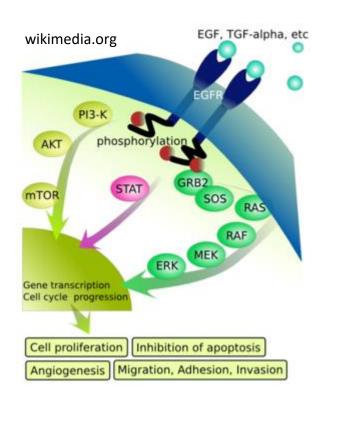
-98 Da loss of phosphoric acid H₃PO₄ during fragmentation



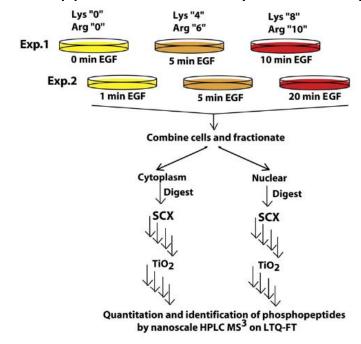
(Threonine changes to 2-aminodehydrobutyric acid, -18 Da)

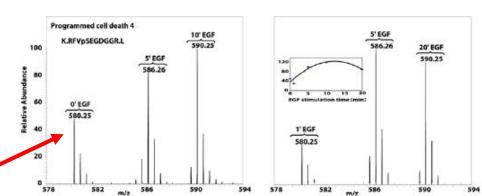
Quantitative Proteomics Reveals Dynamics in Signaling Networks

Phosphorylation dynamics after EGF stimulation



SILAC approach enables dynamic analysis

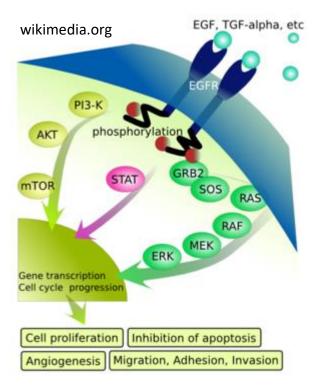


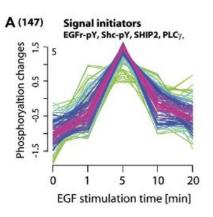


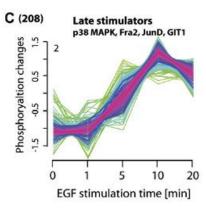
MS spectra triplets

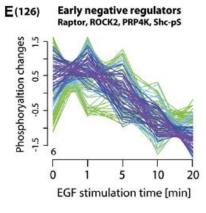
Olsen, et al. Cell, 2006

Phosphorylation dynamics after EGF stimulation









Proteomics & Protein-Protein Interactions

Overview

- Techniques & Technologies
 - Mass Spectrometry
 - Protein-Protein Interactions
 - Quantitative Proteomics
- Applications
 - Representative Studies
- Putting it all together....
 - Databases & Pathways

DNA -> RNA-> PROTEIN



2001

The Sequence of the Human Genome

J. Craig Venter, ^{1*} Mark D. Adams, ¹ Eugene W. Myers, ¹ Peter W. Li, ¹ Richard J. Mural, ¹ Granger G. Sutton, ¹ Hamilton O. Smith, ¹ Mark Yandell, ¹ Cheryl A. Evans, ¹ Robert A. Holt, ¹

articles

Initial sequencing and analysis of the human genome

International Human Genome Sequencing Consortium*



2014

ARTICLE

doi:10.1038/nature13319

Mass-spectrometry-based draft of the human proteome

Mathias Wilhelm¹⁻², Judith Schleg²*, Hannes Hahne¹*, Amin Moghaddas Glolami¹*, Marcus Liebreurd, Midhall M. Savitski², Ennanuel Ziegler', Lars Butzmann, 'Segfried Gesular', Harid Marx', Toby Mathieson', Simone Lemer¹, Karsten Schnathum², Ulf Reimer¹, Holger Wenschuh¹, Martin Mollenhauer¹, Julia Slotta-Huspenina³, Joos-Hendrik Boese², Marcus Bantscheft³, Anja Gersmair², Franz Facher² & Bernhard Kluster⁴.

ARTICLE

doi:10.1038/nature1330

A draft map of the human proteome

Min-Sik Kim^{1,2}, Sneha M. Pinto², Derese Getnet^{4,4}, Raja Sekhar Nirujogi², Srikanth S. Manda², Ragbothama Gherkadyl^{2,2}, Andi K. Madugnud², Dhansshree S. Kelkar², Ruth Isserlin², Shobhi Jain², Joji K. Fhomes², Balylakshmi Muthusamy², Pamela Leal-Rojas^{3,6}, Praveen Kumar³, Nandini A. Sahasrabuddhe², Lawaya Balakrishnan³, Jayshree Advan³, Bijesh George³, Santosh Remse², Lalskimi Dhevi N. Selvan³, Arun H. Patil³, Vishtalakshi Nanjappa³, Aneesha Radharishnan³, Samarisher Prasad³

The Sequence of the Human Genome. PMID: 11181995

Initial sequencing and analysis of the human genome. PMID: 11237011

A draft map of the human proteome. PMID: 24870542

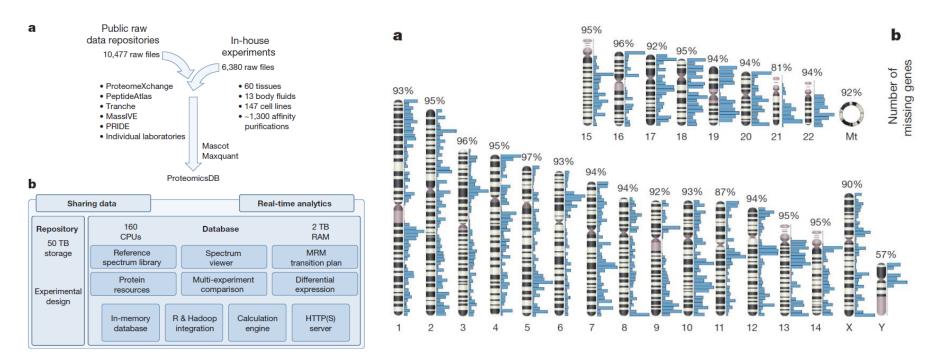
Mass-spectrometry-based draft of the human proteome. PMID: 24870543



Mass-spectrometry-based draft of the human proteome

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- Large Assembly of new and existing data:
- ProteomicsDB, database designed for the real-time analysis of big data https://www.proteomicsdb.org

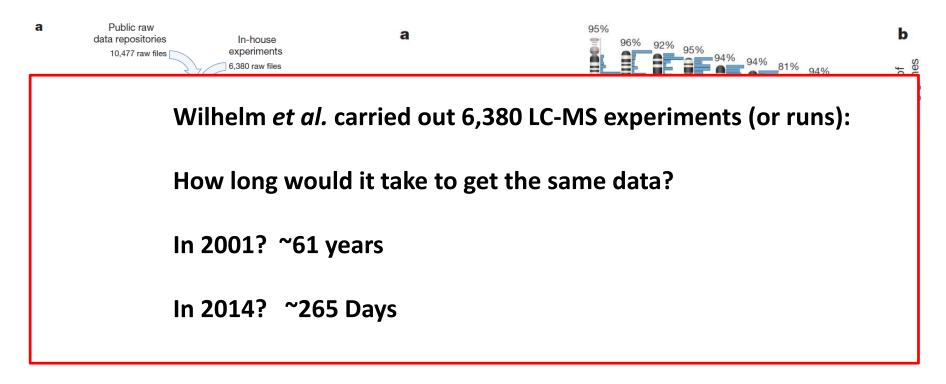




Mass-spectrometry-based draft of the human proteome

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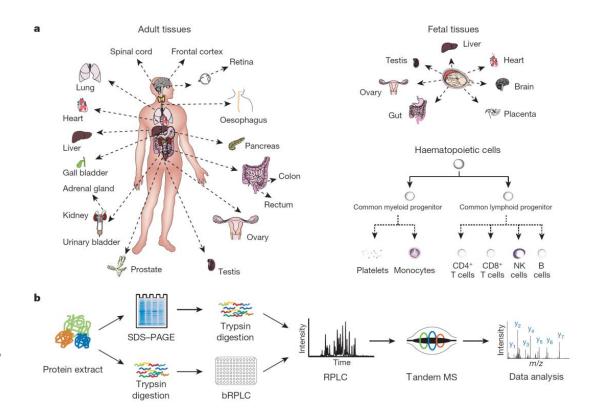




A draft map of the human proteome

Min-Sik Kim^{1,2}, Sneha M. Pinto³, Derese Getnet^{1,4}, Raja Sekhar Nirujogi³, Srikanth S. Manda³, Raghothama Chaerkady^{1,2}, Anil K. Madugundu³, Dhanashree S. Kelkar³, Ruth Isserlin⁵, Shobhit Jain⁵, Joji K. Thomas³, Babylakshmi Muthusamy³, Pamela Leal-Rojas^{1,6}, Praveen Kumar³, Nandini A. Sahasrabuddhe³, Lavanya Balakrishnan³, Jayshree Advani³, Bijesh George³, Santosh Renuse³, Lakshmi Dhevi N. Selvan³, Arun H. Patil³, Vishalakshi Nanjappa³, Anessha Radhakrishnan³, Samarjeet Prasad¹,

- New, large collection of proteomics data
 - 30 histologically normal human samples
 - 17 adult tissues,
 - 7 fetal tissues
 - 6 purified primary haematopoietic cells
- 17,294 genes accounting for approximately 84% of the total annotated protein-coding genes in humans.



Proteomics Databases: Peptide depositories



http://www.peptideatlas.org/builds/

TaxID	Date	Number of Samples	Peptide Inclusion Cutoff	Number of Peptide- Spectrum Matches (PSMs)	Number of Distinct Peptides	Reference Database	Peptide Sequences	Peptide CDS Coordinates	Peptide CDS and Chromosomal Coordinates	Database Tables
9606	Mar 2015	1011	PSM FDR = 0.0002	133,638,335	1,025,698	Ensembl v78+UPSP+Trembl201412+14IPI 3.87+cRAP+nextprotSNP	APD Hs all.fasta	prot map	chrom map	MYSQL,XML

Protein Identification Terminology used in PeptideAtlas http://www.peptideatlas.org/docs/protein_ident_terms.php

- Each PeptideAtlas build is associated with a reference database usually a combination of several protein sequence databases (Swiss-Prot, IPI, Ensembl ...)
- From the reference database, any protein that contains any observed peptide is considered to be a member of the Atlas.
- It is easy to see that the entire list of proteins in an Atlas is going to be highly redundant. Thus, we label each Atlas protein using the terminology below.
 - The term "observed peptides" in this context refers to the set of peptides in the PeptideAtlas build.
 - These peptides are selected using a PSM (peptide spectrum match)

Proteomics Databases: Peptide depositories

http://thegpm.org/GPMDB/index.html



The Global Proteome Machine

Proteomics data analysis, reuse and validation for biological and biomedical research.

The GPMDB Project

gpmDB: Design

gpmDB was designed to be a simplification and extension of the MIAPE scheme proposed by the PSI committee of HUPO. Rather than being a complete record of a proteomics experiment, this database holds the minimum amount of information necessary for certain bioinformatics-related tasks, such as sequence assignment validation. Most of the data is held in a set of XML files: the database serves as an index to those files, allowing for very rapid lookups and reduced database storage requirements. We call this combination of a relational database with XML data XIAPE (Xml Information About a Proteomics Experiment).

The Minimum Information About a Proteomics Experiment (MIAPE)

http://www.psidev.info/node/91

Nature Biotechnology 25, 887 - 893 (2007) PMID: 17687369

Methods Mol Biol. 2014;1072:765-80. PMID: 24136562

Proteomics Databases: Peptide depositories



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About Human Proteome Map

The Human Proteome Map (HPM) portal is an interactive resource to the scientific community by integrating the massive peptide sequencing result from the draft map of the human proteome project. The project was based on LC-MS/MS by utilizing of high resolution and high accuracy Fourier transform mass spectrometry. All mass spectrometry data including precursors and HCD-derived fragments were acquired on the Orbitrap mass analyzers in the high-high mode. Currently, the HPM contains direct evidence of translation of a number of protein products derived from over 17,000 human genes covering >84% of the annotated protein-coding genes in humans based on >290,000 non-redundant peptide identifications of multiple organs/tissues and cell types from individuals with clinically defined healthy tissues. This includes 17 adult tissues, 6 primary hematopoietic cells and 7 fetal tissues. The HPM portal provides an interactive web resource by reorganizing the label-free quantitative proteomic data set in a simple graphical view. In addition, the portal provides selected reaction monitoring (SRM) information for all peptides identified.

Statistics	
Otatistics)	
Organs/cell types	30
Genes identified	17,294
Proteins identified	30,057
Peptide sequences	293,700
N-terminal peptides	4,297
Splice junctional peptides	66,947
Samples	85
Adult tissues	17
Fetal tissues	7
Cell types	6

ARTICLE

doi:10.1038/nature13302

A draft map of the human proteome

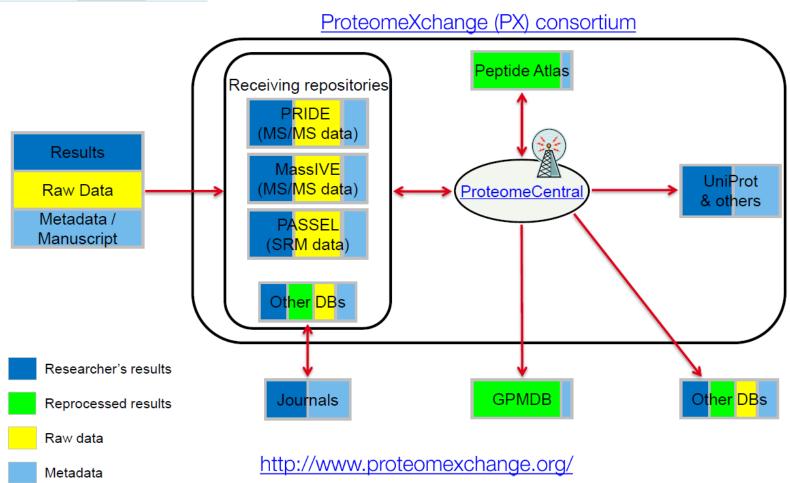
Min-Sik Kim^{1,2}, Sneha M. Pinto³, Derese Getnet^{1,4}, Raja Sekhar Nirujogi³, Srikanth S. Manda³, Raghothama Chaerkady^{1,2}, Anil K. Madugundu³, Dhanashree S. Kelkar³, Ruth Isserlin⁵, Shobhit Jain⁵, Joji K. Thomas³, Babylakshmi Muthusamy³, Pamela Leal-Rojas^{1,6}, Praveen Kumar³, Nandini A. Sahasrabuddhe³, Lavanya Balakrishnan³, Jayshree Advani³, Bijesh George³, Santosh Renuse³, Lakshmi Dhevi N. Selvan³, Arun H. Patil³, Vishalakshi Nanjappa³, Anessha Radhakrishnan³, Samarjeet Prasad¹,

Kim & Akhilesh Pandey et al., *Nature*, 2014. PMID: 24870542

Proteomics Databases: Integrated Resources



http://www.proteomexchange.org/

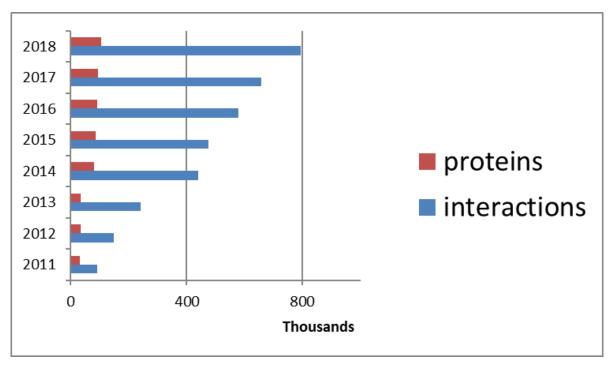


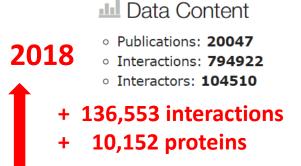
Slide modified from "Computational Mass Spectrometry-Based Proteomics 6th Maxquant Summer School" 21-25 July 2014 Emanuele Alpi, UniProt and PRIDE Development

Protein-Protein Interaction Databases



http://www.ebi.ac.uk/intact/





Data Content

2017 • Publications: 14451 • Interactions: 658369

o Interactors: 94358

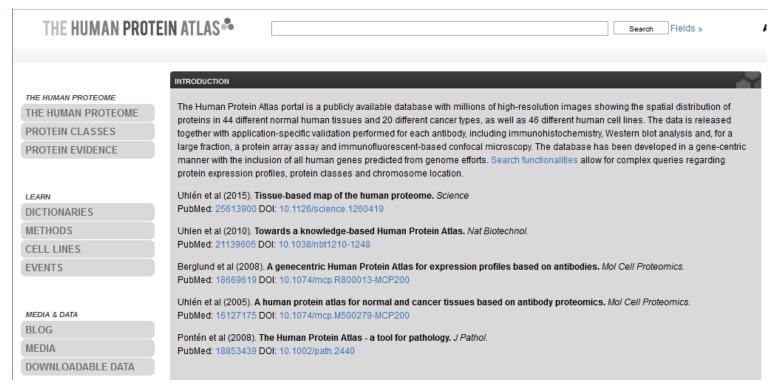
Proteomics Databases: Integrated Resources Beyond Mass Spectrometry

http://www.proteinatlas.org/



A Tissue-Based Map of the Human Proteome

Here, we summarize our current knowledge regarding the human proteome mainly achieved through antibody-based methods combined with transcriptomics analysis across all major tissues and organs of the human body. A large number of lists can be accessed with direct links to gene-specific images of the corresponding proteins in the different tissues and organs. Read more



Proteomics & Protein-Protein Interactions

Jesse Rinehart, PhD
Biomedical Data Science: Mining & Modeling
CBB 752, Spring 2018



