"Mass spectrometry-based proteomics centered multi-omics: applications to clinical lung disease research"

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COMMUNITY-BASED SOFTW ARE

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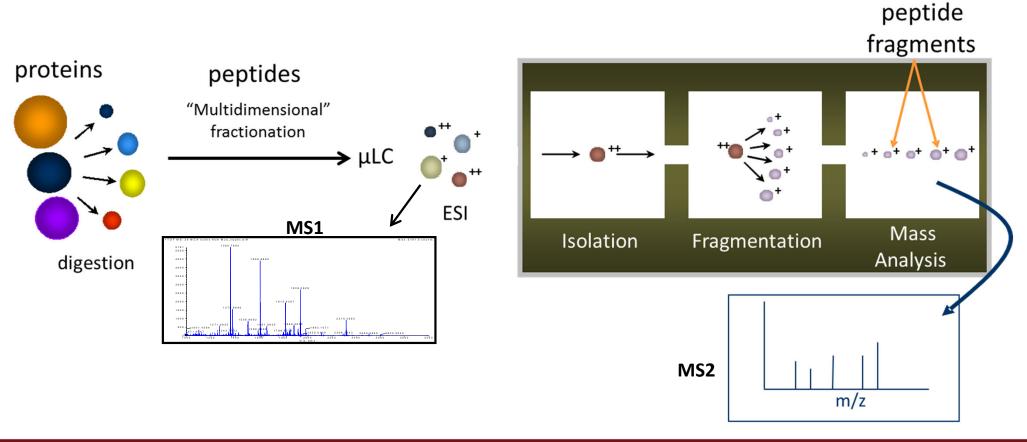
Outline

- Translational mass spectrometry-based proteomics studies of lung disease and health: technology platforms and workflow development
- Application to studies of obstructive lung disease at the intersection of HIV:
 - Quantitative proteomic profiling of bronchoalveolar lavage fluid (BALF)
 - Peptidomics: characterizing protease activity in lung disease
- Metaproteomics: microbial contributors to host disease and biology
 - Value of metaproteomics
 - Challenges and technology platforms
 - Applications to cystic fibrosis



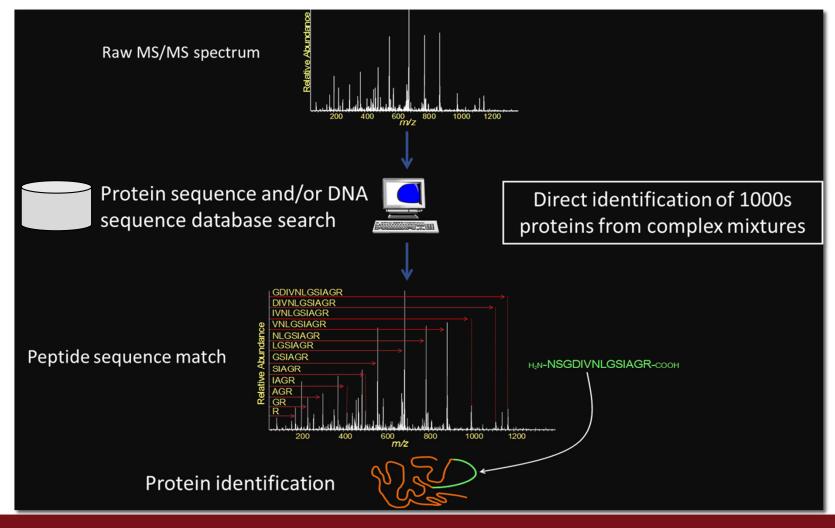
Mass spectrometry-based proteomics: core technology platform

Peptide fractionation coupled to tandem mass spectrometry (MS/MS)





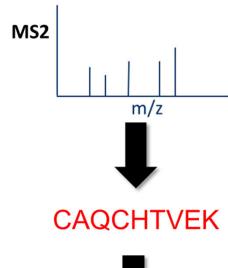
Matching amino acid sequences to MS/MS data







Inferring protein identity for peptide sequences





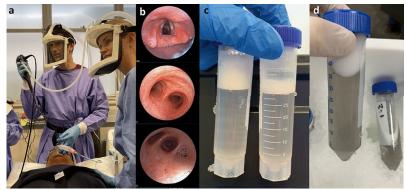
Cytochrome C

NH2GDVEKGKKIFVQKCAQCHTVEKGGKHKTGPNLHGLFGRKTGQAPGFT YTDANKNKGITWKEETLMEYLENPKKYIPGTKMIFAGIKKKTEREDLIAYLK KATNEcooh

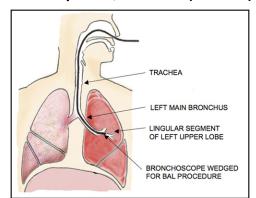




Application to clinical lung samples: bronchoalveolar lavage fluid (BALF)



Sci Rep 13, 8859 (2023)

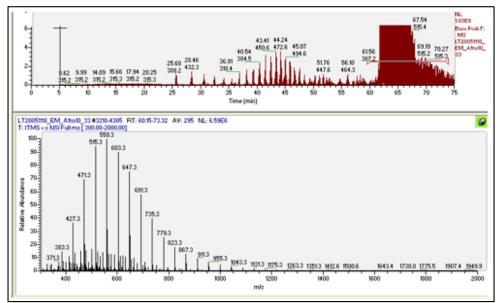


Proteomics Clin Appl. 3, 1044 (2009)



Challenges for MS-based proteomics analysis of BALF

- Dominated by high abundance proteins (primarily from plasma)
- Relatively large volumes of fluid requiring concentration
- Removal of potential contaminants: surfactants, lipids



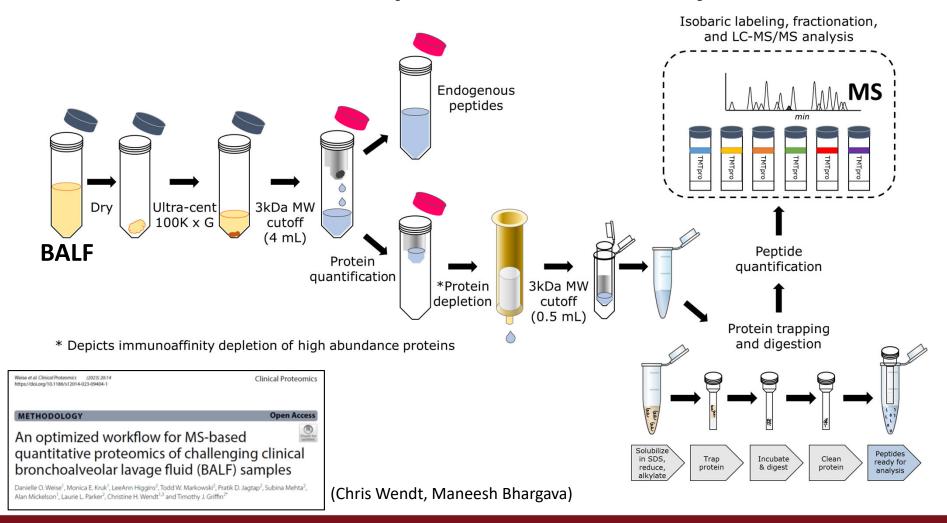




https://www.ohsu.edu/proteomics-shared-resource/sample-preparation

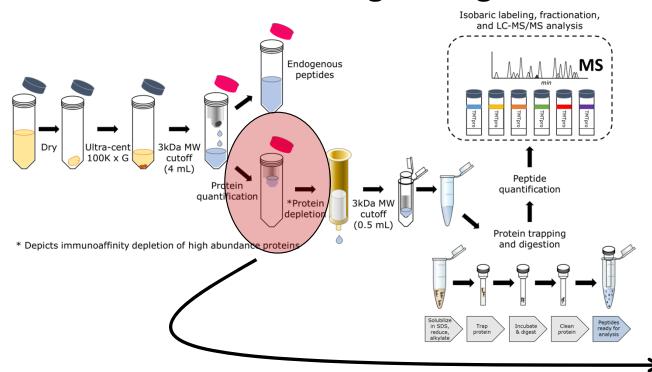


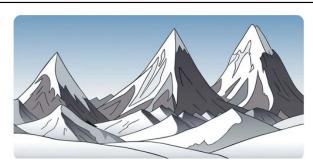
A workflow for MS-based proteomics of BALF: years in the making





Challenge 1: High abundant proteins





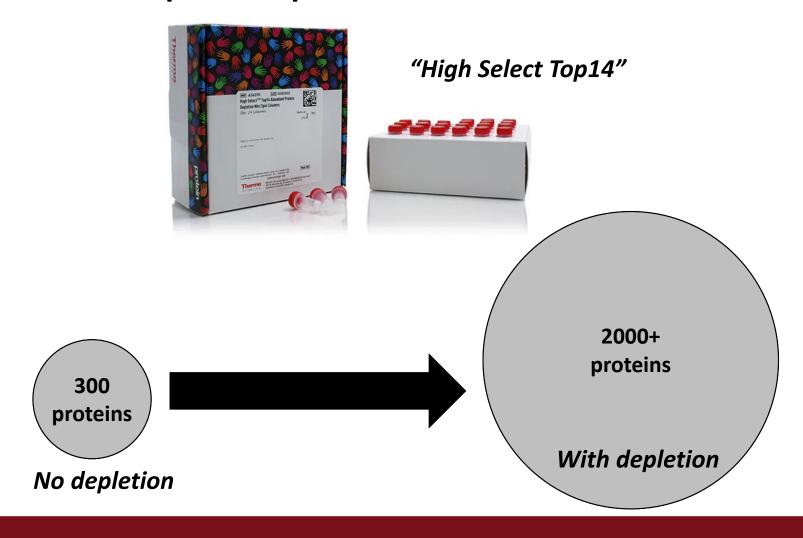
The "high peaks" – plasma proteins:

- serum albumin (90% of total protein?)
- Transferrin
- alpha-1 antitrypsin
- immunoglobulin A and G
 Proteomics Clin Appl. 2014 8:737

Limits protein detection to only a few hundred in BALF

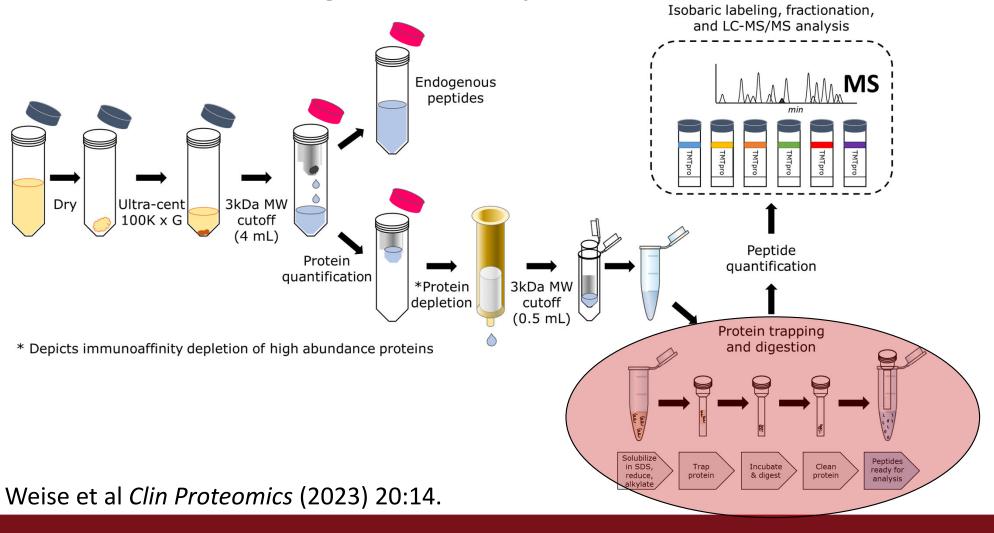


Protein depletion spin columns: works for BALF!



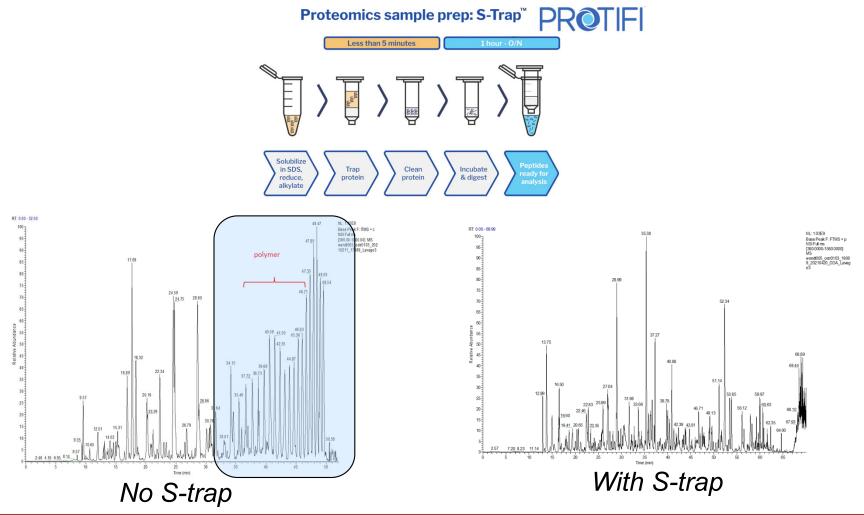


Challenge: MS-incompatible contaminants



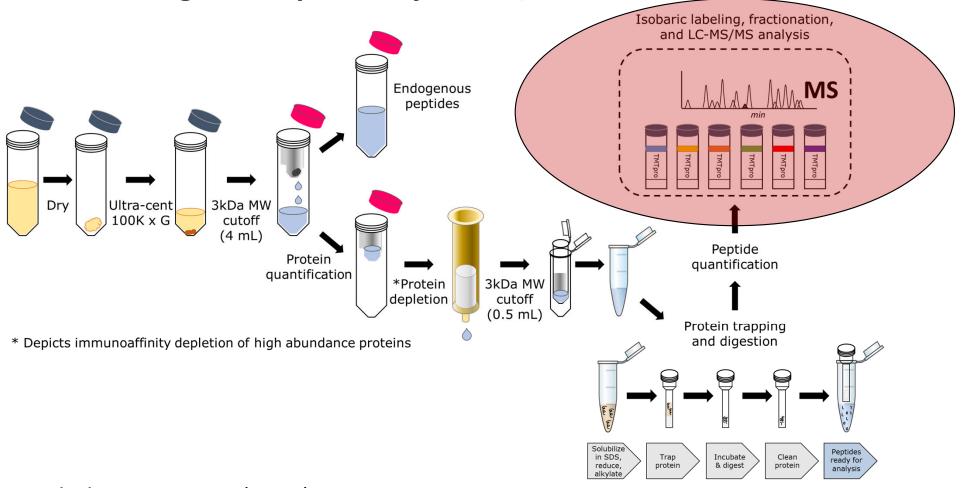


A workflow to meet the challenges of BALF for MS-based protoemics





Challenge: Compatibility with quantitative proteomics



Weise et al Clin Proteomics (2023) 20:14.



Challenge: Compatibility with quantitative proteomics

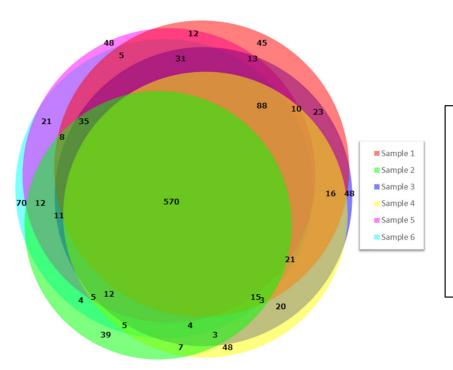


Table 2 Results from a TMT-based quantitative repeatability experiment in BALF samples using our sample preparation workflow

| Peptides identifed (1% FDR) | Protein groups (1% FDR) | Average replicate TMT abundance ratios | | |
|--------------------------------|-------------------------|--|--|--|
| 20,399 | 1816 | 0.92 ± 0.27 (29.6% CV) | | |

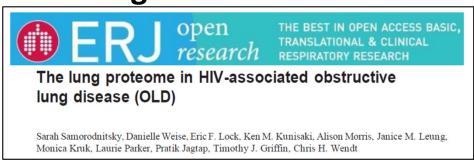
Protein identifications across 6 separate BALF samples – pilot study



Weise et al Clin Proteomics (2023) 20:14.



Putting the workflow to use



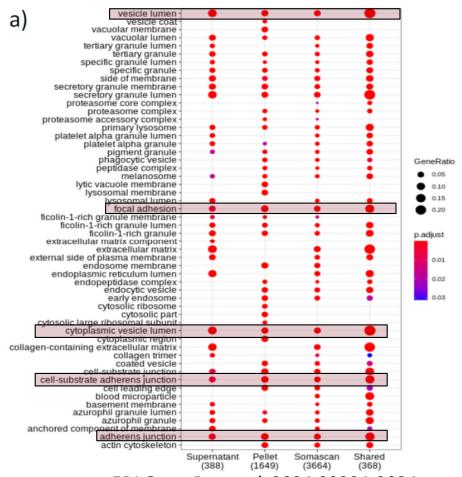
- Quantitative proteomics in BALF samples from persons living with HIV with OLD (n=26) and without OLD (n=26)
- Pearson correlation tests to identify associations between each protein and lung function (FEV1pp)
- Goal: Identify pathways and/or protein markers predictive of OLD development

Summary of results

- 3800 proteins identified by MS (cellular and soluble BALF)
- 254 proteins that correlated with FEV1pp controlling for smoking status



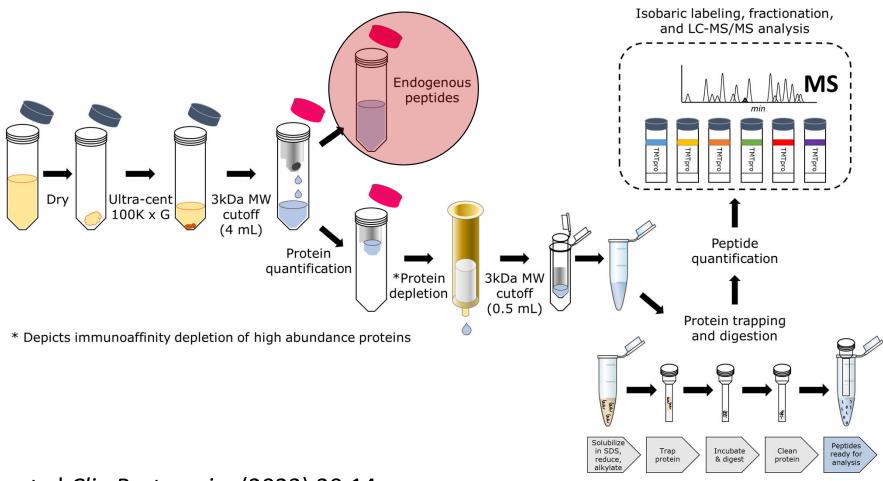
Outcome: Cell adhesion proteins as potential predictor of OLD



ERJ Open Research 2024 00204-2024



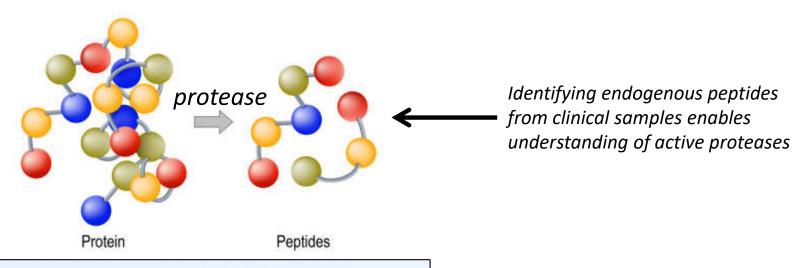
Application: Peptidomic analysis of BALF



Weise et al Clin Proteomics (2023) 20:14.



Value of peptidomics in clinical samples



Role of Proteases in Inflammatory Lung Diseases

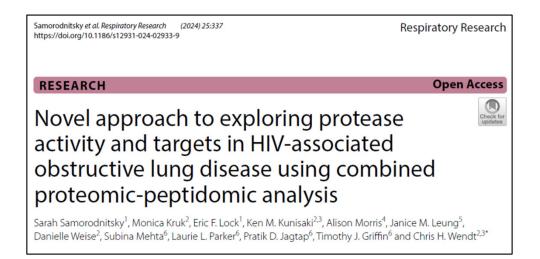
Chapter | First Online: 13 November 2013 pp 361–385 | Cite this chapter

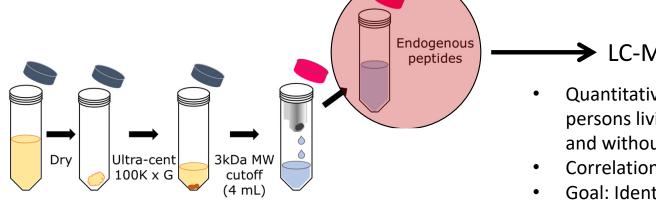
Kaustav Chakraborty & Arindam Bhattacharyya

Part of the book series: Advances in Biochemistry in Health and Disease ((ABHD, volume 7))









➤ LC-MS/MS

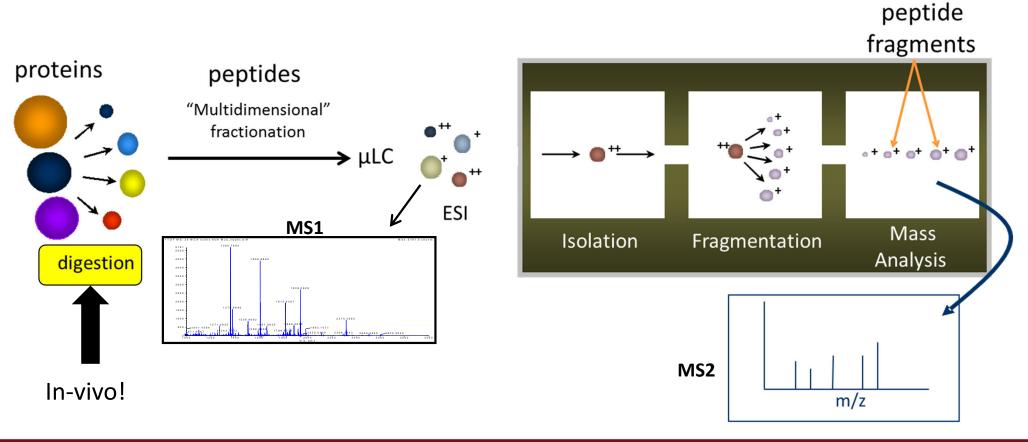
- Quantitative peptidomics in BALF samples from persons living with HIV (PLWH) with OLD (n=25) and without OLD (n=26)
- Correlation to measured lung function (FEV1pp)
- Goal: Identify enriched protease activities associated with OLD in PLWH populations





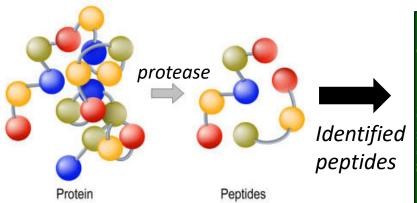
Mass spectrometry-based proteomics: core technology platform

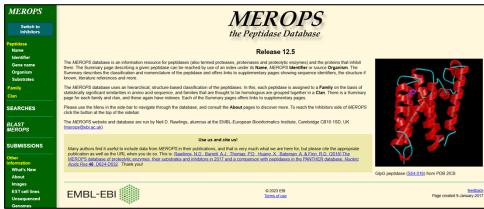
Peptide fractionation coupled to tandem mass spectrometry (MS/MS)





Bioinformatics tools to identify enriched proteases





| | Protein Description | Peptide Sequence | Cleave | Protease | z-score |
|--------------------|--|------------------|-----------|----------|---------|
| Known Associations | Collagen alpha-2(I) chain | QSLQEETVRKGPAG | GPAG DRGP | MMP13 | 2.77 |
| | Tubulin beta chain | AILVDLEPGTMDSVR | YVPR AILV | GZMK | 2.67 |
| | Histone H1.4 | AAAGYDVEK | KKAL AAAG | GZMM | 2.20 |
| | Histone H2A type 2-C | AGLQFPVGR | RSSR AGLQ | GZMA | 2.06 |
| | Ezrin | LRAKEELERQAV | RMAA LRAK | MMP3 | 1.95 |
| | Actin, cytoplasmic 1 | APEEHPVLL | PVLL TEAP | CTSG | 1.80 |
| | Kininogen-1 | SSRIGEIKEETT | SPFR SSRI | KLK1 | 1.81 |
| | Phosphoglycerate kinase 1 | SLEPVAVELK | VELK SLLG | ELANE | 1.74 |
| - | Vimentin | NLRETNLD | TNLD SLPL | CASP6 | 1.61 |
| | Macrophage-capping protein | QVLGPKPALK | PALK EGNP | CTSS | 1.57 |
| Associations | Ubiquitin-60S ribosomal protein L40 | IIEPSLR | LRGG IIEP | USP4 | 3.84 |
| | Calreticulin | KSGTIFDNF | FDNF LITN | CTSA | 2.99 |
| | Collagen alpha-2(I) chain | QSLQEETVRKGPAG | GPAG DRGP | MMP13 | 2.77 |
| cia | Prothrombin | ESYIDGR | IDGR IVEG | F10 | 2.58 |
| SSO | Bactericidal permeability-increasing protein | NVVLQPHQNF | HQNF LLFG | CTSA | 2.58 |
| | Alpha-1-antitrypsin | LSGVTEEAPLK | APLK LSKA | CPA4 | 2.55 |
| Unknown | Deleted in malignant brain tumors 1 protein (I2) | DVGSYQEKVDV | RSKR DVGS | FURIN | 2.51 |
| | Apolipoprotein A-IV | EKESQDKTL | STFK EKES | ATG4B | 2.50 |
| | Natural resistance-associated macrophage protein 1 | MTGDKGPQR | GPQR LSGS | MMP8 | 2.30 |
| | Putative neutrophil cytosol factor 1B | GDTFIRH | M GDTF | METAP1D | 2.25 |





Monica Kruk



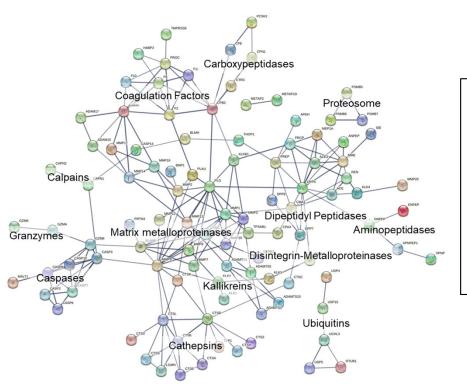
Results: enriched proteases based on analysis of peptidomics data

protease and FEV1pp.

PRTN3

ATG4B

 Combined analysis of quantified proteases from BALF with analysis of endogenous peptide sequences resulting from protease action



| Processor and Pr | | | | | | | |
|--|------------------------------|--------------|---------|-------|--|--|--|
| Gene | Protease | No. Proteins | p-Value | FDR | | | |
| ELANE | Neutrophil elastase | 23 | 0.003 | 0.018 | | | |
| GZMM | Granzyme M | 20 | 0.245 | 0.420 | | | |
| CTSD | Cathepsin D | 19 | 0.233 | 0.412 | | | |
| CTSE | Cathepsin E | 19 | 0.154 | 0.330 | | | |
| ASPRV1 | Aspartic peptidase | 18 | 0.166 | 0.346 | | | |
| BLMH | Bleomycin hydrolase | 18 | 0.405 | 0.594 | | | |
| CELA1 | Chymotrypsin like elastase 1 | 18 | 0.000 | 0.005 | | | |
| CTSG | Cathepsin G | 18 | 0.001 | 0.010 | | | |

Table 5 Top 10 most active proteases mapped to cleaved proteins. P-value and FDR describes significance of correlation between

22 proteases correlated to FEV1pp function

Autophagy related 4B cysteine peptidase

Proteases are ordered by number of associated proteins cleaved

Network view



0.000

0.890

0.000

0.785

Using (meta)proteomics to characterize microbial contributors to lung disease and health



Current Research in Immunology
Volume 4, 2023, 100065



Lung microbiome and origins of the respiratory diseases

José Belizário ^a 🌣 ☒ , Miguel Garay-Malpartida ^a, Joel Faintuch ^b

Commensal and pathogenic microbes are key contributors to lung health and disease







MICROBIOME





METAGENOMICS

TAXONOMY

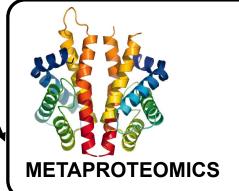
function

Potential to unravel the mechanistic details of microbial interactions with host / environment by analyzing the functional dynamics

of the microbiome.



TAXONOMY function



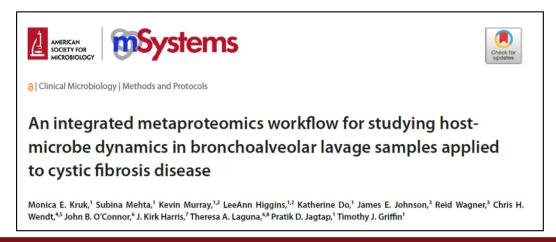
TAXONOMY FUNCTION

(Slide courtesy of Pratik Jagtap)



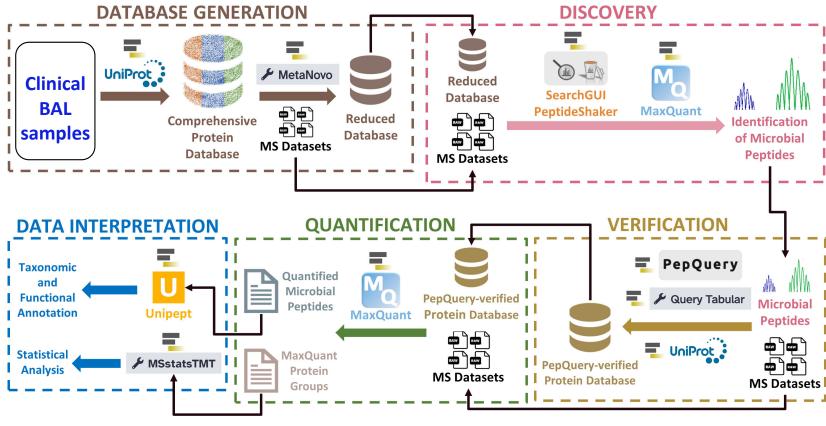
Challenges of metaproteomics in clinical human samples (e.g. BALF)

- Detecting relatively low abundance microbial proteins from abundant host (human) proteins
- Knowledge of which microbe proteomes might be present (solved by multiomics using metagenomic information as a guide)
- Matching MS/MS to very large metaproteomic protein sequence databases (10M+ sequences)
- Mapping identified microbial peptides to specific taxa and annotating protein function





A clinical metaproteomics bionformatic workflow

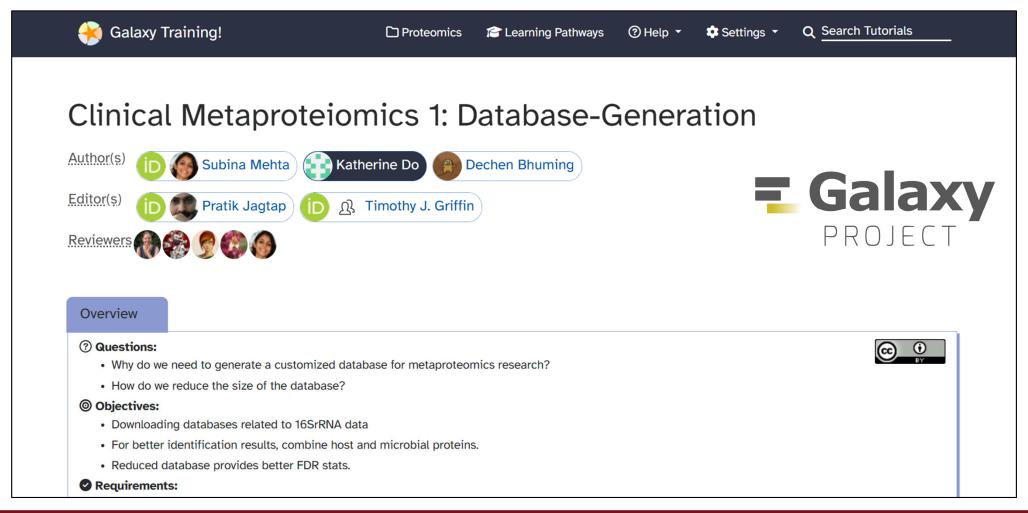




mSphere (2024) 9:e0079323. (Pratik Jagtap, Galaxy-P team)

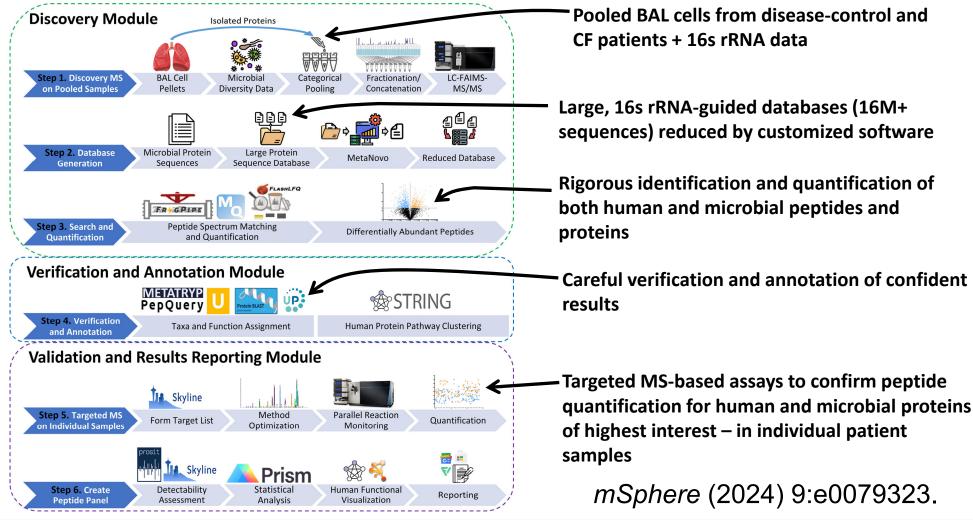


Developing accessible software tools for multi-omics analysis





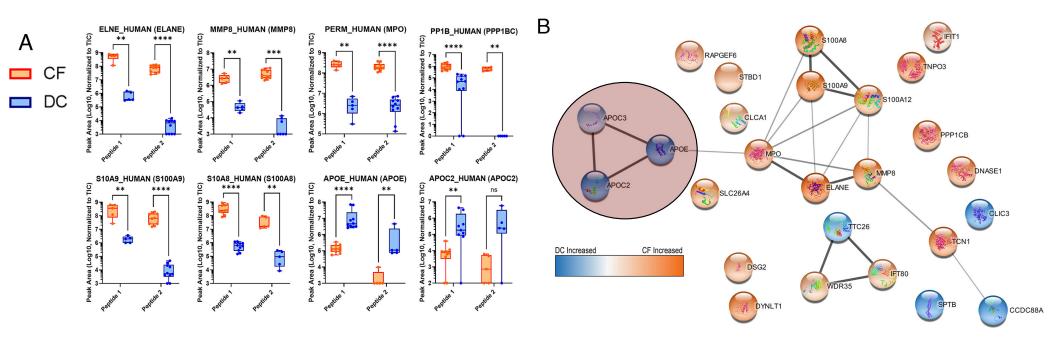
Application to microbial contributors in Cystic Fibrosis





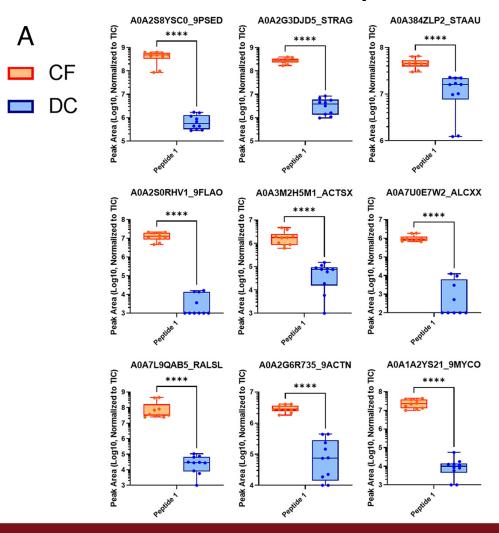
Results: Human proteins

- Discovered from quantitative proteomics data in pooled samples
- Validated and quantified in individual patient samples





Results: Bacterial proteins



Peptides detected from:

Streptococcus Pseudomonas Mycobacterium Ralstonia



Outcomes of CF metaproteomics in BAL samples

- Demonstration of combined protein-level analysis of human and microbial contributors to CF
- Demonstration of targeted MS-based assays to characterize individual CF patient samples
- Opens possibility of clinical assays for simultaneous characterization of host-microbial protein markers that may indicate disease state and/or monitor therapy outcomes



What's next: improving the state of proteomics and multi-omics

Improved instrumentation: 10X faster and 10X more sensitive?





• Improved methods to integrated 'omic levels of information for better molecular understanding of disease in clinical samples (more predictive, more personalized)



Integration of other MS-based information (metabolomics)



