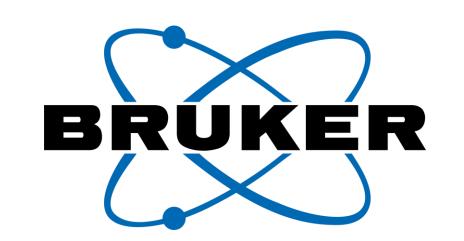
TIMSrescore: timsTOF-optimized PSM rescoring boosts identification rates for immunopeptidomics

Jonathan R. Krieger¹, Tharan Srikumar¹, Dennis Trede², George Rosenberger³, Arthur Declercq⁴, Ralf Gabriels⁴, Robbin Bouwmeester⁴, Sven Degroeve⁴, Michele Genangeli², Lennart Martins⁴







Introduction:

Bruker ProteoScape (BPS) it a GPU-powered platform delivering parallel computing capabilities and real-time database search results for bottom-up proteomics. Real-time database searching alleviates the data analysis bottleneck by providing both quality control as well as experimental results immediately after measurement, reducing the wait for entire projects to be completed before starting analysis.

BPS represents a platform of various softwares to search dia-PASEF, diagonal-PASEF, and dda-PASEF data with using various community accepted search algorithms such as Spectronaut® 19 and ProLuCID. Herein we introduce the integration of the MS2 Rescore

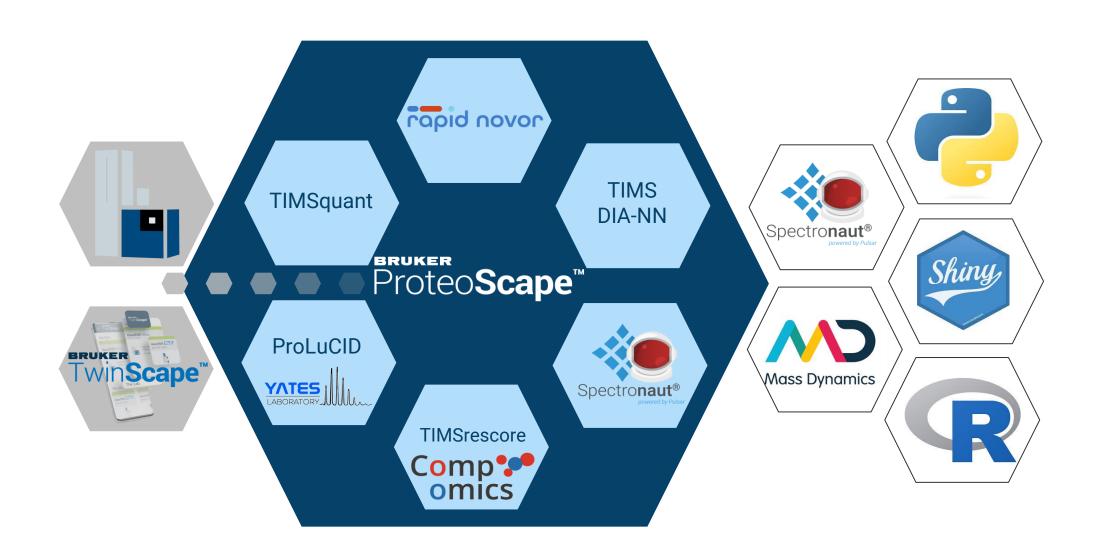


Fig. 1: An overview of the Bruker ProteoScape (BPS) platform. The platform enables researches to search several modes of timsTOF data in real time. Data can then be exported for post processing through several different applications.

Methods

The peptide fragmentation (MS2PIP) and collisional cross section (IM2Deep) predictors employed in TIMSrescore have been trained using a diverse set of timsTOF dda-PASEF PSMs. The updated models are available as part of MS2Rescore and can be utilized independently from Bruker ProteoScape (BPS). The complete workflow in BPS is illustrated in Fig 2. For evaluation we used 3 datasets. 1) Replicate injections of HeLa protein lysate digested with elastase. 2) Replicate phospho-enriched data from mouse cell lines with or without treatment with LPA provided by the laboratory of Prof. Stanely Stevens Jr. from a pilot project. The project was eventually analyzed by dia-PASEF and published https://doi.org/10.3389/fonc.2023.1048419. 3) The RCC tumor samples from Hoenisch et al., 2023. were also reprocessed with our pipeline. Representing triplicate measurements of HLA class I and II enriched samples.

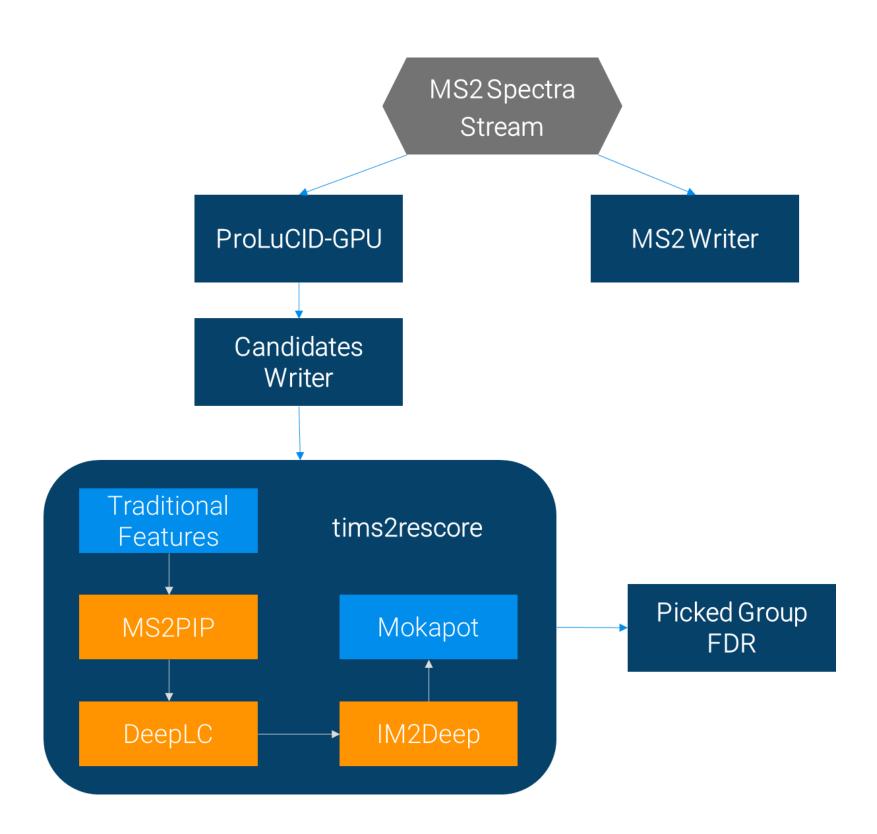


Fig. 2: The TIMSrescore workflow. As MS2 spectra are acquired from any timsTOF series instrument, data is streamed to ProteoScape. User defined search parameters are used by ProLuCID-GPU to generate a list of candidates for each spectra. Once acquisition ends, the candidates are written to parquet file and passed to the tims²rescore module. tims²rescore module adds many additional vectors to the traditional features provided by ProLuCID, including comparisons with predicted fragment ion intensities (via MS2PIP), retention time (via DeepLC) and CCS (via IM2Deep). This aggregated feature map is sent to Mokapot for PSM and peptide validation. The validated PSM list is then processed by picked group FDR.

Hoenisch Gravel, N. *et al.* TOFIMS mass spectrometry-based immunopeptidomics refines tumor antigen identification. *Nat Commun* 14, 7472 (2023).

Declercq, A. et al. MS2Rescore: Data-Driven Rescoring Dramatically Boosts Immunopeptide Identification Rates. Molecular & Cellular Proteomics 21, (2022).

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Results:

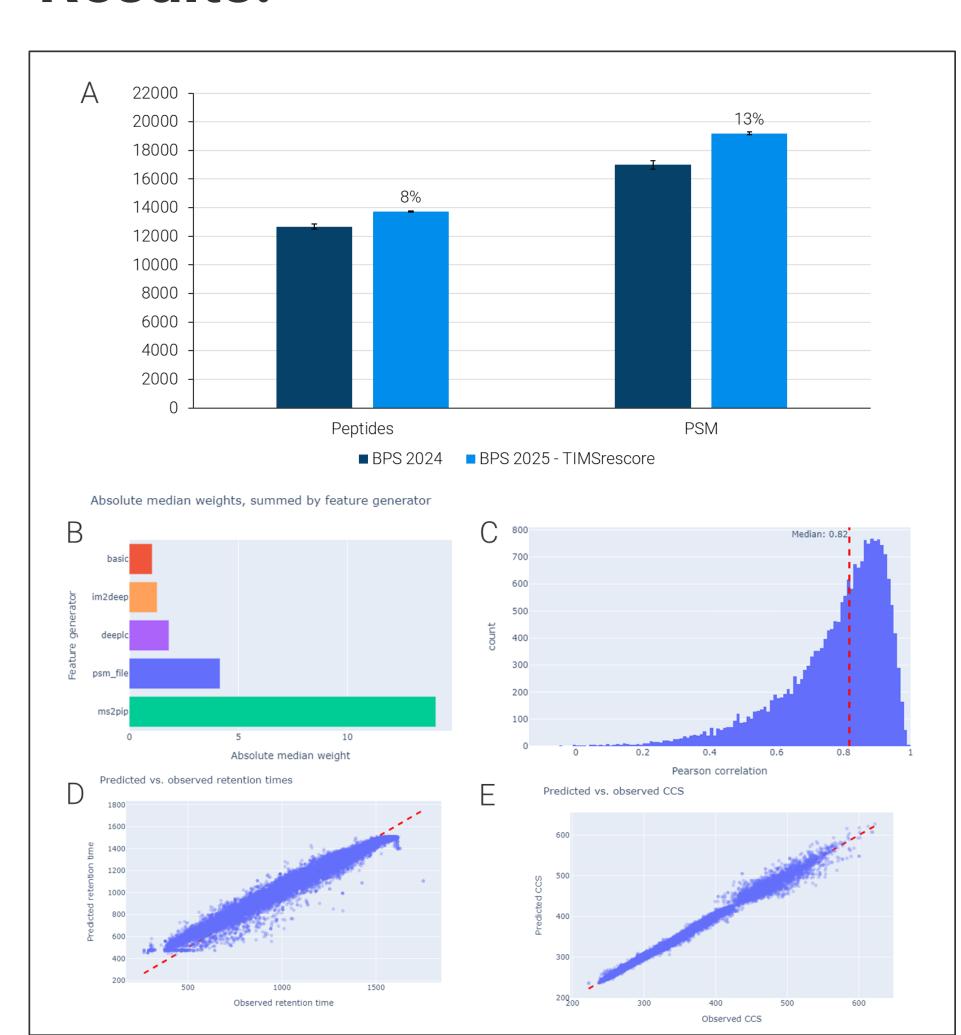


Fig. 3: TIMSrescore workflow in BPS 2025 increases confidently identified peptides by >8% and PSM by 13%. 5ng of HeLa cell lysates digested with elastase were analyzed with 22min active LC gradient on timsTOF Ultra in triplicate. (A) The average number of peptides and PSMs are shown with and without rescoring. Feature usage in TIMSrescore. (B) Absolute median weights summed by feature generator. Each feature generator contributes to the overall rescoring process of separating target and decoy PSMs. (C) MS2PIP distribution of Pearson coefficients for all target PSMs (D) Deep-LC, and (E) IM2Deep model performance showing predicted vs observed RT and CCS.

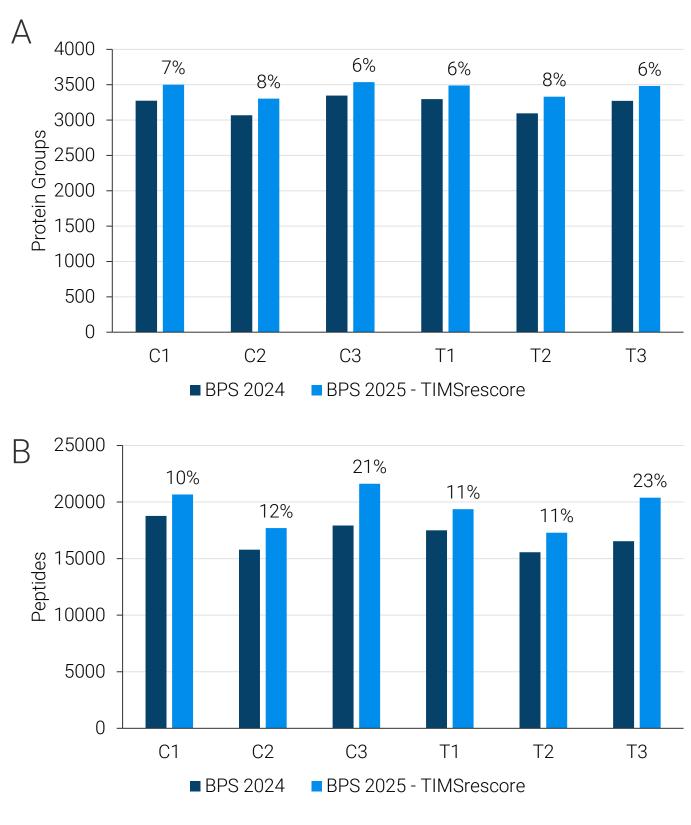


Fig. 4: Biological triplicates of control and treated phospho-enriched samples processed with and without TIMSrescore. TIMSrescore was able to increase the number of confidently identified (A) protein groups by \sim 7% (on average) and (B) peptides by \sim 15% (on average).

Optimizing the MS2rescore feature generators for timsTOF data was critical for the improvements provided by the workflow. Details of the optimization will be provided in a manuscript that is in preparation.

In the HeLa elastase dataset, TIMSrescore increased peptide identification by 8% and PSMs by 13%. For the phosphoproteomics dataset, TIMSrescore increased recovery of peptides 10-23% and protein groups by 6-8%. For the immunoproteomic dataset, a considerable increase of 21% could be observed, indicating that rescoring is particularly impactful for large search spaces.

For all the datasets, the timsTOF optimizations of the fragmentation predictors were critical. For example, in the MHC-I dataset it led to an improvement of the median Pearson correlation to 0.88 from 0.53 (standard MS2PIP HCD model). IM2Deep was created based on principles of DeepLC, allowing for the CCS prediction for modified peptides, even if the modifications wasn't observed in the training data. The retention time predictions and ion mobility predictions had varying levels of (positive) contribution depending on the dataset.

Finally, we show that with the newly designed and released diagonal-PASEF modes, specifically midia-PASEF, TIMSrescore provides a significant boost in protein group identifications,

Conflict of Interest Statement

JK, TS, DT, GR, & MG are current employees of Bruker Daltonics

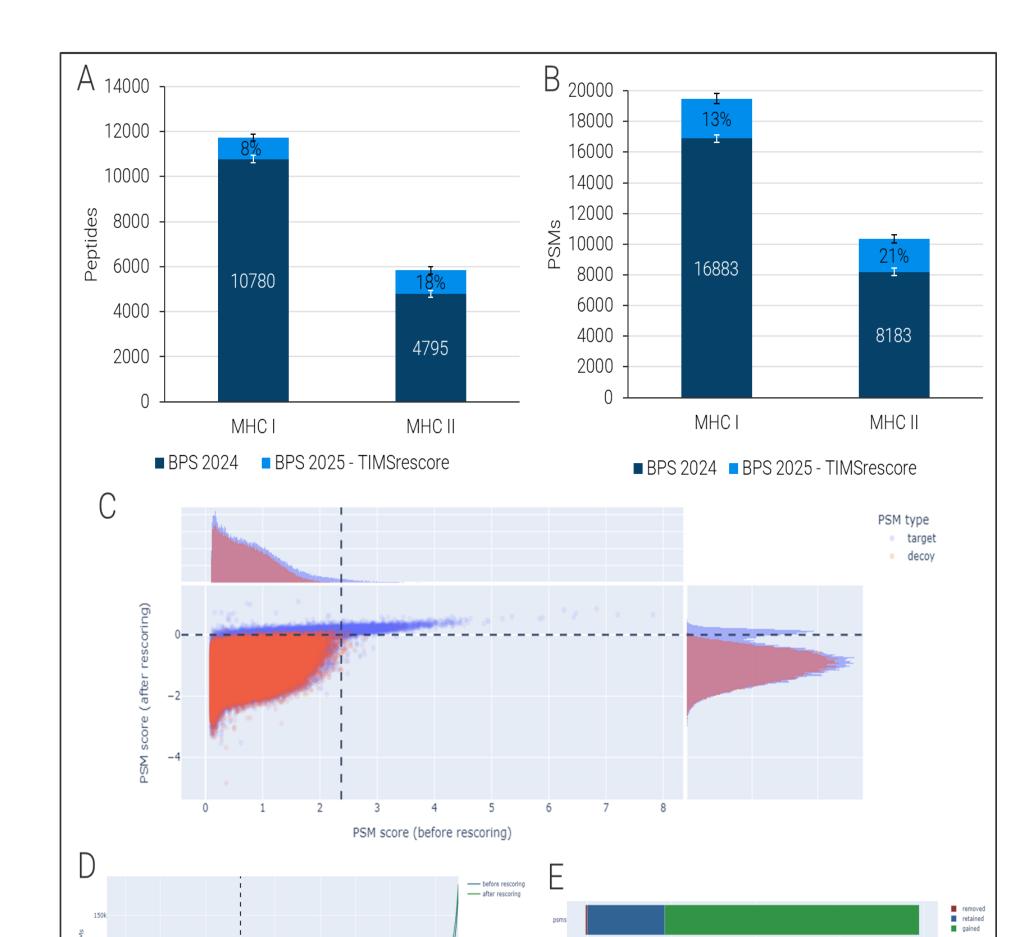


Fig 5: TIMSrescore for Immunopeptidomics. Peptide and PSM identifications with and without rescoring for both MHC-I (n=3) and MHC-II datasets (n=3). TIMSrescore increased confident MHC-I peptides by 8% and MHC-II peptides by 18%. Similarly, TIMSrescore increased MHC-I PSMs by 13% and MHC-II PSMs by 21% versus the standard workflow available in BPS for dda-PASEF analysis. (C) Scatterplot of Score Comparison for the MHC-I dataset. Target (blue) and decoy (red) PSMs before rescoring are shown on the x-axis and after rescoring are shown on the y-axis. The upper left quadrant are the PSMs only identified after rescoring. (D) False Discovery Rate Comparison. This shows the number of identified target PSMs in function of the FDR threshold. (E)Identification overlap showing PSMs, and peptides removed, retained and gained by the rescoring engine.

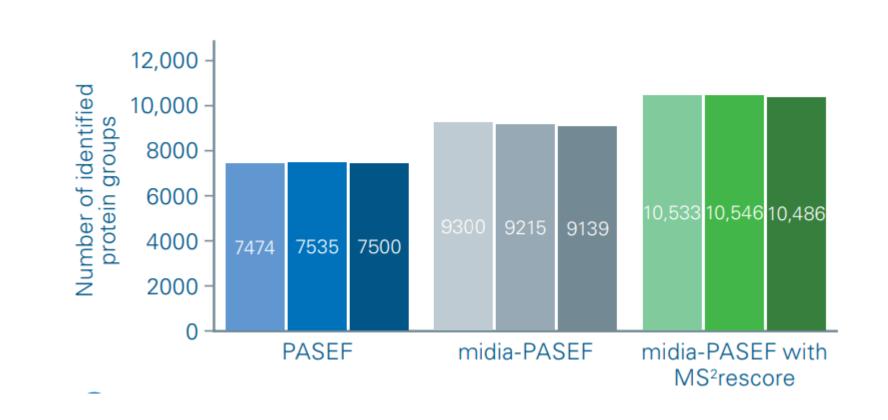


Fig 6: Use of Rescore with midia-PASEF data. Here we show both the increase in identifications of using midia-PASEF over traditional dda-PASEF but then further the boost that TIMSrescore gives to such data. This is the first application of TIMSrescore being applied to data acquired in the data-independent acquisition space.

Conclusions:

- TIMSrescore represents a timsTOFoptimized rescoring approach that can improve recovery of peptide identifications
- •TIMSrescore increased identifications in all tested datasets with an average increase of 6-20%.
- TIMSrescore makes use of all available dimensions of timsTOF data including fragmentation, retention time, and the TIMS dimension
- ■TIMSrescore can be applied to the new midia-PASEF based acquisition modes