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Introduction

To provide a comprehensive background for the following chapters that focus on the interaction of human mesenchymal stromal cells (hMSCs) with multiple myeloma (MM) cells, this

Aims

This project defines these aims:

- Characterize the interaction between myeloma cells and mesenchymal stromal cells
- Develop methods
- Face the challenge of time-dependent cell adhesion through
- Provide tools to analyze multidimensional cell adhesion data

Summarising Discussion

Time-Lapse Microscopy Added Intuition to Exploratory Cell Biology

- When starting this project, dissemination has not been the main topic. - Surprisingly, Time-lapse identified detaching cells - Hence, Time-lapse proved pivotal for this project, shifting the focus onto in vitro dissemination.

Microscopy holds vast amounts of information. Cell movements themselves add a lot more info. Time-lapse video has proven invaluable for exploratory cell biology

the most important key insight on the mechanism of dissemination identified by timelapse was Cell Division - further insights were multiple time measurements

measuring the minimum time for detachments to begin, or the time required for daughter cells to re-attach to the hMSC monolayer. Such mechanistic insights

Other methods like RNAseq and survival analysis did provide molecular and clinical connections, time-lapse microscopy documented cell interactions as-is, but allowed for a deep and intuitive understanding of cryptic molecular data, placing the conclusions into a context that answer key questions about potential and limits of this study, such as the aggregation behavior of INA-6 cells.

Novel Methods of Isolating Adhering Subpopulations

In this work, innovative *in vitro* methodologies (Well Plate Sandwich Centrifugation and V-Well adhesion Assay) were developed. this was required to fill in gaps of isolating cells with minimized variability introduced by user-bias to clearly separate subpopulations and precisely quantify them.

cite all those methods for cell isolation! - Turning around wellplates: Doesn't allow isolation, just quantification - The author did not show all his washing experiments - Washing is very bad (data not shown): Highly dependent on user: position of cell on well bottom (border cells receive less force), position of pipette tip in well (depth, angle and position on bottom) - This motivated us to explore more reproducible methods

It's a challenge: either quantify cell population, or isolate them! - It's better to specialize in one method, than to do both poorly - Well Plate Sandwich Centrifugation is badly suited for quantification, but possible - we switched to developing V-well adhesion assay for quantification - We realized, V-well isolation allows both ultra precise quantification and isolation of small amounts of cells! - unmatched precision through centrifugation, no washing - But V-well pellets comprise only few cells requiring a lot of technical replicates and an untiring pipetting hand

The Well Plate Sandwich Centrifugation (WPSC) used two different techniques to dissociate MA-INA6 cells from the hMSC monolayer. This had no impact on the ratio of isolated MA-INA6 to nMA-INA6, since nMA-INA6 isolation was performed prior to dissociation using the same protocol consistently. We tried this to test if MACS was really necessary, after all it is costly, time-consuming, introduces an antibody bias and requires cell cold-treatment during antibody: Strong pipetting ('Wash') and repeated Accutase treatment followed by magnetic activated cell sorting ('MACS').

Outlook: High-Value Research Topics for Myeloma Research

As an Outlook, the Author lists research topics arising from this work that have great potential for breakthroughs in myeloma research.

Cell Division as a Mechanism for Dissemination Initiation: The author describes how the detachment of daughter cells from the mother cell after a cycle of hMSC-(re)attachment and proliferation could be a key mechanism in myeloma dissemination. This mechanism was shown in other studies of extravasation. The author sees great potential in this mechanism as a target for future research. It is probably under-researched due to requirement of sophisticated time-lapse equipment, yet the simplicity of detachment through cell division is intriguing through its simplicity. It implies asymmetric cell division. Cancer cells are known to divide asymmetrically, e.g. moving miRNAs to one daughter cell.

Time as a Key Parameter: The area Thermodynamics of started with scientists measuring how long it takes for gases to cool down. The author claims, by measuring the time it takes for cancer cells to detach could lead to breakthroughs in research of myeloma dissemination.

- Cell adhesion is highly time-dependent. - Cell detachment is required for metastasis and dissemination -

key mechanistic insights

measuring the minimum time for detachments to begin, or the time required for daughter cells to re-attach to the hMSC monolayer. Such mechanistic insights

The author recommends high time resolutions, e.g. 1 frame every 15 min, which is a high res-

olution for common live cell imaging when compared to Purschke et al. (2010). Time-resolution was mostly limited by available disk space. Investing into more hard drives is worth it, since

Lists of Adhesion Gene Associated With Prolonged Patient Survival: The author lists adhesion genes that are associated with prolonged patient survival. These genes are highly expressed in myeloma samples from patients with longer overall

Semi-Automation was Critical for Establishing in vitro Methods

In vitro research is valued for their speed at creating precise data (Moleiro et al., 2017). In this work, the development and publication of innovative in vitro methodologies necessitated the adoption of semi-automated data analysis tools. These novel methods introduced complexities that span multiple experimental parameters, making the results multidimensional. This demanded precise, efficient and standardized data handling capabilities which were facilitated by Python tools like seaborn and plotastic.

Inherent Multidimensionality of Adhesion Studies: Cell adhesion studies often involve multiple independent parameters, posing significant analytical challenges. Two critical dimensions are particularly notable: 'Subpopulation' and 'Time'. Analyzing cell adhesion often involves isolation of adherent and non-adherent subpopulations, effectively introducing 'Subpopulation' as a vital dimension in the dataset (Dziadowicz et al., 2022). This study specifically categorized cells into three levels of MSC-interaction: CM-INA6, nMA-INA6, and MA-INA6. Furthermore, the dynamic nature of cell adhesion processes is profoundly influenced by the factor 'Time', making it a crucial experimental parameter for investigation (Rebl et al., 2010; McKay et al., 1997; Bolado-Carrancio et al., 2020). This work includes extensive time-lapse microscopy experiments utilizing a high time resolution (1 frame every 15 min), similar to those time resolutions used by Purschke et al. (2010). This precision was required for key mechanistic insights on hMSC-INA-6 interaction dynamics. These included identifying rolling movements of nMA-INA6 daughter cells around MA-INA6 mother cells, measuring the minimum time for INA-6 detachments to begin, and measuring the time required for daughter cells to re-attach to the hMSC monolayer, etc. Next to mechanistic insights, adhesion time played a crucial methodological role in this study as well: During the V-Well adhesion experiments, we did not know initially how long INA-6 cells required to form strong adhesion with hMSCs before pelleting nMA-INA6, but required a timepoint with hour precision to capture detachments after cell division that was accelerated through prior cell cycle synchronization at M-Phase.

The extensive facetting features of seaborn and plotastic were essential for visualizing these multidimensional datasets, allowing for quick exploration of the data (Waskom, 2021).

Further Contributions and Remedies to Multidimensional Complexity: In addi-

tion to 'Subpopulation' and 'Time', this study faced additional layers of complexity that were managed through semi-automated analysis.

Experiments typically involved at least three biological replicates and corresponding technical replicates. Although replicates were not treated as independent variables – instead serving for displaying variance—they can add substantially to the data management workload. In this work, semi-automation nullified the manual burdens of handling replicates: pandas was used to automate aggregation of technical replicates into means after removing technical outliers thorugh z-score thresholding, while the jupyter notebook format allowed for reviewing filtered data for specific data losses. The removal of technical noise was especially relevant for qPCR data, where low gene expression can lead to sudden increase in Ct value (non-detects). In fact, the decision to remove or impute non-detects is under active discussion, however, available algorithms are hard to understand for non bioinformaticians, but also do not separate biological from technical variance, which is considered bad practice by Motulsky (2018) (McCall et al., 2014; Sherina, 2020). Semi-automation also nullified the burden of handling biological replicates: The automatic aggregation of datapoints during plotting is a key feature of seaborn, on which plotastic was built. Without such automation, calculating means and standard deviations for simple barplots would have required extensive manual computation in *Microsoft* Excel, or tedious plot configurations in Graphpad Prism due to limited facetting functionality of multiple variable tables (GraphPad Prism 10 User Guide, 2024).

Replicates can expand datasets as this factor comprises a lot of levels. Similarly, the factor 'Gene' multiplied the dataset by a total of 30 genes when validating RNAseq data with RT-qPCR. With three subpopulations, one timepoint, eleven biological replicates, and three technical replicates, the qPCR dataset used in this study grew to 2970 datapoints to be statistically analyzed and visualized. This is a manageable size for manual analysis, but the effort involved illustrates the definition of semi-big data.

Methodological variability also introduced additional dimensions: The Well Plate Sandwich Centrifugation (WPSC) used two different techniques to dissociate MA-INA6 cells from the hMSC monolayer: Strong pipetting ('Wash') and repeated Accutase treatment followed by magnetic activated cell sorting ('MACS'). These variations, recorded as the factor 'method', further complicated the dataset. Although this distinction is not discussed in this work – rather pooled into one group—, this showcases how protocol changes can add dimensions to the dataset that are not necessarily relevant for the biological question but essential for method optimizations and validation.

Agility During Establishment of V-Well Assay: The concept of agility in laboratory research, inspired by the Agile Manifesto's principle of "Responding to change over following a plan" (Manifesto for Agile Software Development, 2001), is increasingly relevant in biomedical

research (West, 2018; Quanbeck et al., 2022). This adaptive approach was particularly crucial during the development of the V-Well adhesion assay in this study. Semi-automation using python significantly enhanced this agility, allowing rapid statistical testing and visualization of data, which would have taken considerably longer if done manually. This capability enabled real-time adjustments to the experimental technique during live microscopy sessions, integrating raw data tables directly into Python scripts for immediate analysis. Such an agile and adaptive work environment, facilitated by python tools and seaborn, proved invaluable for refining the *in vitro* methods being developed. Additionally, the simplicity offered by seaborn for complex data plotting, such as the cell cycle profiling shown in Appendix A.1: Fig. 3, which required minimal code to produce a detailed series of 24 histograms, underscores the utility of semi-automation in enhancing laboratory efficiency. While this work does not quantify the full benefits of semi-automation, the author's experiences suggest significant potential impacts on the speed and adaptability of method development in biomedical research.

Re-Designing, Re-Analysis and Re-Annotation with plotastic: Establishing new methods of *in vitro* dissemination required not just innovative experimental protocols, but also adaptive ways to visually present complex data. This need for adaptability is crucial during the publication process, where researchers must often experiment with different ways to visually represent their findings to best convey their significance. This process typically involves frequent adjustments to how data is displayed in plots—changes that can be cumbersome with traditional tools, even reaching the limits of Python packages like *seaborn*.

plotastic addresses these challenges by not only automating statistics, but to enhance the flexibility of data visualization as well, making it easier to modify how data is presented without repetitive manual adjustments. Its key design feature is the centralized storage of plot parameters (y, x, hue, row, col), such as which data points are shown on the x-axis, what categories are highlighted by color (hue), and how data is grouped into separate plots (by columns and/or rows) into separate plots. This centralization means that once these parameters are set, they not only automate statistical analysis, but also can be automatically applied across all subsequent adjustments made to the plot. These subsequent adjustment include four tasks that can be inefficient for repeated re-design using Python alone:

- Plotting multiple layers
- Re-arranging facets
- Statistical annotations
- Fine-Tuning for publication quality

multiple plots and statistical analyses without needing to redefine them each time. This

contrasts with seaborn, where changing these parameters typically requires adjusting each subsequent edits applied to the plot, such as fine-tuning the visual representation or adding statistical annotations using additional code with matplotlib and/or statannotations.

For instance, plotastic's .switch() method allows researchers to easily shift the arrangement of plots – for example, switching the data represented on the x-axis with that represented by color – to explore different perspectives of the data quickly. This is particularly useful when trying to find the most effective way to illustrate complex interactions or trends that might not be immediately apparent.

plotastic simplifies the creation of publication-quality figures by automating style adjustments that are typically manually coded with matplotlib when using seaborn. These include adjustments like angled x-axis labels or consistent visual styles across multiple figures, which are important for maintaining the professional appearance of published data.

Moreover, plotastic extends the functionality of seaborn by automating the inclusion of statistical annotations directly into plots. This is a significant enhancement because it ensures that any statistical significance noted in the data is immediately visible and correctly updated whenever the data presentation is changed. This feature is invaluable during the peer review process, where researchers may need to adjust their figures to meet reviewer demands or clarify their findings. For this work in particular, reviewer comments required extensive restructuring of Chapter 1 Fig. 4 and 5 during the publication of Kuric et al. (2024).

a task that can be labor-intensive with tools like *GraphPad Prism*, which require manual re-annotation and recalculating of statistics with each change.

was pivotal during the

By significantly reducing the manual effort required to reconfigure and reanalyze data visualizations, plotastic not only saves time but also enhances the clarity and communicative power of research findings. This makes it an essential tool for researchers, particularly those developing new methodologies in fields like myeloma dissemination, where clear and accurate data presentation is critical for advancing understanding and treatment strategies.

Intermediate Conclusion: Together, adhesion studies, especially combined with methodologies that further add dimensions (Cell Cycle profiling), explode experimental complexity, which require sophisticated tools for high requirements posed by multidimensionality.

Combining this with the speed of the in vitro feedback loop of acquiring quick results, evaluating results, adjusting the protocol, try again, repeat.

These requirements were facilitated by seaborn for quick processing of intermediate results during method optimizations and by plotastic for designing publication-grade analyses and

figures. Developing plotastic was mostly a necessity that arose from late-stage data processing (statistics, and publication figures), filling seaborn's shortcomings, but also ensured standards for consistent plot formatting and statistical analysis. The author lays emphasis on plotastic as an infrastructure for quick and confident re-structuring of publication figures that include statistical annotations, especially during his peer-review process.

Challenges of Integrating plotastic into Biomedicine

Although plotastic is designed for the overall scientific community and has passed peer-review, time will tell if the single author's vision of an optimized automated statistical workflow can be extrapolated to biomedicine. The author's himself continues to use plotastic in future projects that require visualizing single datapoints and statistical rigor.

For biomedicine overall, the author sees the greatest challenge in adopting plotastic since it is a tool based on a command line interface (CLI), whereas the majority of biologists prefer graphical user interfaces (GUI). The author argues that a CLI can perform everything that a GUI can, but better, faster, and more efficiently, provided that one is willing to undergo the switch, which can be intimidating. However, many python tools already provide extremely easy commands (e.g. seaborn), and plotastic further lowers the barrier to entry. The author also argues that ChatGPT is a strong argument to switch to CLI, since large language models are highly compatible with text, which is the main format of CLIs. This allows for understanding context, providing code drafts, identifying errors, adding and changing analyses, autocompletion of repetitive commands, follow-up questions and many more.

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Appendices

A Supplementary Data & Methods

A.1 Figures

A.2 Tables

A.3 Materials & Methods

B Documentation of plotastic