

Development of Molecular Dynamics Workflow and Dataset for Assessing Pharmacoperone Responsiveness in α -Galactosidase A Mutants

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Abstract

This document is a **technical report** accompanying an open repository that provides a reproducible set of procedures, simulation data, and analysis scripts related to the molecular dynamics (MD) characterization of α -galactosidase A (α GAL) mutants relevant to Fabry disease. Our objective is to establish and document a workflow for deriving MD-based structural and dynamical descriptors that may inform future models of pharmacoperone responsiveness to migalastat. To this end, we have built and simulated wild-type and mutant α GAL structures in both apo and holo forms, including explicit glycosylation and ligand binding, using a standardized preparation protocol and multi-replica, microsecond-scale MD simulations. The pipeline standardizes structure preparation (including glycosylation and ligand binding), multi-replica simulations, and basic RMSD/RMSF-based analyses, enabling downstream use in predictive modeling. This report is a resource release aimed at providing (i) a well-defined, reproducible protocol for similar studies, (ii) a curated dataset for downstream machine-learning and structural analyses, and (iii) a baseline for future extensions to additional mutants, descriptors, and predictive modeling efforts.

1 Introduction

Fabry disease is a multi-systemic, X-linked, lysosomal storage disease caused by decreased activity of α -galactosidase A (α GAL), an enzyme that catalyses the removal of the terminal α -galactose residue from polysaccharides, glycolipids and glycopeptides. The defect of α GAL causes lysosomal accumulations of neutral glycosphingolipids and globotriaosylceramide GL-3 resulting in chronic pain, vascular degeneration, cardiac abnormalities and other symptoms, which make it difficult to have an early diagnosis [1]. The phenotype impairment depends on the residual amount of the enzyme's activity; in fact, the disease

is associated with hundreds of point mutation, the majority of which being missense mutations on the GLA gene (Xq22.1), but also deletions and insertions, resulting in different intensity of symptoms. These mutations generate partially unfolded proteins that can have lower to zero activity or can be unstable and directly lead to degradation [2].

Currently two specific treatments are available [3]:

1. Intravenous enzyme replacement therapies, consisting in infusion of functional α GAL or recombinant proteins with the same functions.
2. Pharmacological chaperone oral therapy with migalastat (Figure 3), a small molecule that allows the recovery or the keeping of the folded structure.

Migalastat treatment is the preferred approach because it is less invasive; however it still presents several issues: (i) chaperone therapy does not provide a definitive cure, and patients must continue treatment throughout their lives; (ii) not all mutations, whether known or unknown, can be addressed by this therapy; and (iii) in certain situations, such as in children under 12 years of age, during pregnancy, or while breastfeeding, its use is not recommended.

Regarding point (ii), chaperone therapy facilitates the correct folding of partially misfolded proteins. For this reason, mutations that strongly impair the folding cannot be recovered. Mutation databases reporting migalastat amenability are available at the galafold amenability website [4], to facilitate the treatment choice, but need constant update to evaluate newly discovered mutations associated to Fabry disease symptoms.

This report documents and releases a reproducible molecular dynamics (MD) workflow and associated dataset for α GAL pharmacoperone amenability studies. It describes: (i) system preparation and standardization procedures, (ii) simulation protocols and data structure, and (iii) example descriptive analyses (RMSD, RMSF, ligand residence times). The purpose is to provide a technical foundation for further research rather than to draw mechanistic conclusions.

2 Approach

The objective of the present work is to evaluate whether it is possible to derive a set of descriptors from MD simulations, a computational technique that captures the time-dependent behavior of biomolecules at atomic resolution, to complement migalastat amenability predictions.

We developed a computational pipeline to prepare, simulate and test α GAL protein in wild-type and mutant form both in presence (holo) and absence (apo) of migalastat with the final goal to derive additional predictors to be used in machine-learning models of responsiveness from MD simulations, taking in account flexibility, stability and other features.

1–50	LDNGLARTPTMGWLHWERFCNLDCCQEEPDCISEKLFMEMAELMVSEGW
51–100	KDAGYEYLCIDDCWMA PQRDSEGR LQADPQRFPHGIRQLANYVHSGGLKL
101–150	GIYADVGNKTCAGFPGSFGYYDIDAQTFADWGVDLLKFDGCYCDSLENLA
151–200	DGYKHMSLALNRTGRSIVYSCEWPLYMWPFQKPNYTEIRQYCNHWRNFAD
201–250	IDDSWKSISILDWTSFNQERIVDVAGPGGWNDPDM LVIGNFGLSWNQV
251–300	TQMALWAIMAAPLFMSNDLRHISPAKALLQDKDVIAINQDPLGKQGYQL
301–350	RQGDNFEVWERPLSGLAWAVAMINRQEIGGPRSYTIAVASLGKGVACNPA
351–398	CFITQLLPVKRKLGFYEWTSRLRSHINPTGTVLLQLENTMQMSLKDLL

Table 1: Sequence of each α GAL monomer simulated (from PDB ID: 3GXT). Residues glycosylated in the wild-type protein are marked in red color.

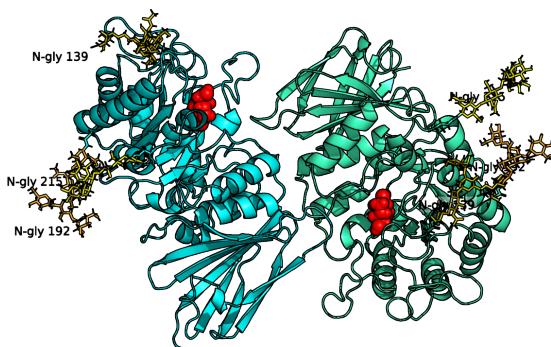


Figure 1: α GAL post-translational modifications. The two subunits are respectively in cyan and aquamarine. We substitute the original glycans with those shown in Figure 2, here shown in shades of yellow, paired by residue position, and replaced the original ligand code NOJ, a migalastat homologue, with the actual migalastat structure (DGJ), in red.

2.1 Protein structure

α GAL is a homo-dimeric protein where each subunit is composed by two domains: an N-terminal region including the active site and a C-terminal domain. Structures of the dimer are widely available at RCSB.org under different conditions and in presence of various substrates. For the purpose of this project we selected the PDB ID model 3GXT, α GAL complexed with 1-deoxygalactonijirimycin, a stereochemical variant of migalastat [5].

2.2 Protein glycosylation

Each α GAL monomer contains three glycosylation sites, namely N139, N192 and N215. In particular, N139 and N215 are important for the protein solubility, whereas N192 improves protein secretion [6]. Considering the important role

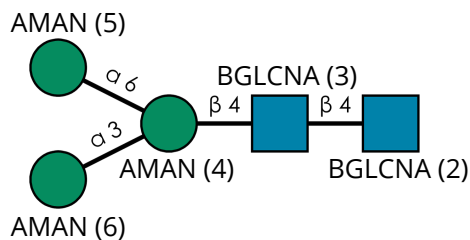


Figure 2: Final glycan structure selected for the analysis (glycan ID: G0026MO). The blue square represent N-acetylglucosamine (BGLCNA), the green circle represent mannose (BMAN). N-glycosylation occurs at the N-acetylglucoamine site. Number in parenthesis indicate the “resid” numbers adopted in the simulated structures.

these glycosylations sites hold for the protein correct functionality, we decided to maintain them in the MD simulations but, as there is no clear information regarding the complete glycan structure for each site, and it would be extremely time consuming to have it in its entirety when running a MD simulation, we decided to maintain only the core structure of the glycan and to keep it identical between all the six sites. In particular, we selected the glycan structure shown at Figure 2, one of the most common human N-glycan cores.

2.3 Migalastat

Migalastat (1-Deoxygalactonojirimycin, here abbreviated as DGJ, Figure 3), is a small iminosugar analog of the terminal galactose moiety of globotriaosylceramide (GL-3), a natural substrate of α GAL. It reversibly binds the active site of the protein thereby facilitating its correct folding and trafficking. Once the lysosome is reached, the acidic pH facilitates migalastat dissociation, allowing the interaction of α GAL with its physiological substrate [7].

2.4 Mutants selection

Fabry diseases is associated to a long list of mutations, often causing different phenotypes, the majority of which is missense or nonsense mutation. The intensity of the symptoms strongly depends on the residual activity of the mutated protein, which depends on the position of the mutation itself and on the kind of residue substitution. In this project we selected two different mutants to start with, namely N215S and R301Q, along with the wild type structure that we consider as our control. This initial set of mutant was chosen both because of the already ongoing *in vitro* studies on their reglycosylation and because they were already present in literature [8, 9, 10, 11, 12, 13].

The substitution N215S is located on the surface of the protein and, in our model, is a N-glycosylation site, which means that when the mutation is present, the glycan is no longer attached to the site leaving the dimer with four

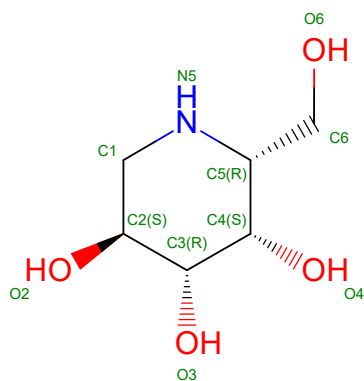


Figure 3: Chemical structure of migalastat (DGJ).

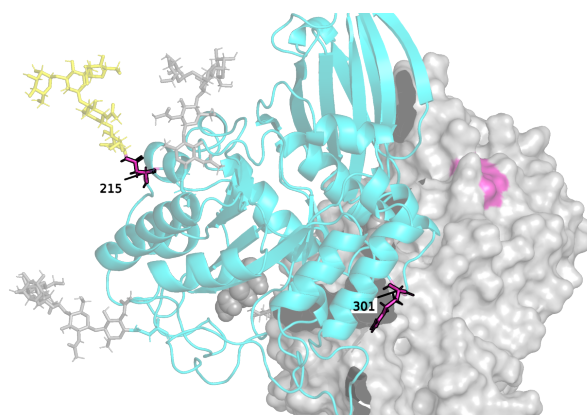


Figure 4: Placement of the mutated residues studied in this project (in magenta) showed on one monomer (chain A, cyan). The mutations are symmetrically applied on both monomers. Residue 215 is also a N-glycosylation site, so when the mutation is applied the glycan (yellow) is no longer attached to the protein.

glycosylations in total instead of six. From the literature, this mutation seems associated to late-onset (variant or non classic phenotype) and low activity of α GAL and resulted amenable by migalastat [8, 9, 10]. R301Q is also located on the surface of the protein, on a loop region at the interface between the monomers. From literature, the mutation is often labelled as cardiac variant of the Fabry disease, with low activity, and migalastat amenable [11, 12, 13].

3 Methods

3.1 Re-glycosylation of the protein

The various α GAL PDB structures available report different glycan structures, due to a partial deglycosylation step included in the crystallisation process. To standardize the systems, we remove the original glycan structures and substitute them with a standard glycan-core structure composed by two N-acetylglucosamines and three mannoses, as shown in Figure 2. We used the online tool glycoshape.org/reglyco on the PDB ID 3GXT structure and the G0026MO (Man(α 1-3)[Man(α 1-6)]Man(β 1-4)GlcNAc(β 1-4)GlcNAc) glycan structure on all the N-glycosylated sites identified by the tool. Once the new structure was generated, it is used as a starting point to produce all the mutants both in apo and holo form.

3.2 System building and preparation

From this point on, all the procedures applied have been collected in a GitHub repository, <https://GitHub.com/giorginolab/aGAL-MD>. In particular, it includes all the required steps to build a system and run a MD simulation:

1. Run the `mol_prep_Fabry.ipynb` in the functions folder of the repository, which allows to build the system by correctly binding glycans to their N-glycosylation sites, insert the desired mutation, provide apo and holo mutants of the same structure, select the length of the simulation.
2. Equilibrate the system using HTMD-provided functions.
3. Run the `production_prep.py` in the functions folder of the repository.
4. Run the production part of the simulation using HTMD-provided functions.
5. (Optional) generate a filtered system by removing water and skipping every 10th frame for faster analysis using `filter_prod.py`.
6. Perform initial analysis of RMSD and RMSF by running the `evaluation.ipynb`.

Further details on how to use the repository are better explained README file present on GitHub.

3.3 Modeling, parameterization, and run conditions

In this project, we evaluated six structures described at Table 2 generated following the previously described steps. The systems were modelled with the CHARMM36 all-atom forcefield [14]. Glycan parameters and topologies were taken from CHARMM36’s `toppar_all36_carb_glycopeptide.str` file. Migalastat was parameterized with the Charmm General Forcefield (program version 4.0) using the ParamChem online service [15], with maximum penalties of 9.600 for parameters and 10.262 for charges. Following the `production_prep.py`, the systems are prepared at pH 7 (at which migalastat binds the protein), solvated with TIP3 water, and ionized with sodium and potassium ions at a concentration of 0.15 M using the software and protocol described in [16].

The equilibration is run in NPT (constant pressure) conditions, at 300 K and for 50 ns, after 1000 steps of minimisation. The production setup is the same except for the NVT (constant volume) conditions and the length of the simulation, 1 μ s each. All of the simulation parameters are available in the simulation scripts. For each condition investigated, we ran three replicas to increase sampling and estimate its statistical uncertainty.

Structure	Replicas	Individual run length (μ s)
apo (wild-type)	3	1
apo_N215S	3	1
apo_R301Q	3	1
DGJ (wild-type holo)	3	1
DGJ_N215S	3	1
DGJ_R301Q	3	1

Table 2: Simulated structures and replicas used in this work.

3.4 Post-simulation analysis

Once the simulations ended, we performed a general evaluation as part of the methods pipeline. This section documents analysis steps rather than interpretation. First, we checked the structures in VMD [17], then we run some analysis based on Root Mean Square Deviation (RMSD) and Root Mean Square Fluctuation (RMSF) to evaluate the protein’s stability thorough the simulation, the glycans and, when present, the ligand. The main scopes of this step are (i) to ensure that the dynamics is stable and the structure did not break, and (ii) to evaluate the glycans and ligand behaviour in the simulated time.

Both RMSD and RMSF are computed with the MoleculeKit projection package [18], in particular using *MetricRmsd* and *MetricFluctuation* classes. In case of RMSF, we computed the square root of the mean value of fluctuation for each residue, saving the results as CSV data tables. The RMSD and RMSF computation part of the analysis is included in the `evaluation.ipynb` notebook.

4 Results and dataset

4.1 Visual check of molecular dynamics simulation

This section documents technical verification steps ensuring simulations behaved as expected, not a biological interpretation of the results. After checking that the simulations ended correctly (i.e., if run with ACEMD [19] under the conditions described in the repository, the last output file should report a "Simulation completed!" line), we proceeded to visualise the structures in the Visual Molecular Dynamics (VMD) software to make sure everything behaved as intended. We note a shared pattern of movement between the different structures and replicas, as in the majority of cases, independently from the presence or absence of a mutation, the monomers rotate pivoting at the dimer interface. In system DGJ_N215S_3 this motion occurs to the extreme causing the monomers partial separation.

4.2 Data reading and interpretation

For a quantitative analysis of the simulations, we provided a notebook, `evaluation.ipynb`, which includes RMSD and RMSF calculation, data storage and plotting of some examples we presented in this report. All the outputs, tables and plots, of the notebook are present on our GitHub repository in the results folder, which includes:

- *tables* folder, to store the numeric data computed either calling the `rmsd` or `rmsf` functions, in form of comma-separated values (CSV) tables.
- *plots* folder, to collect the graphical representation of tables data.

4.3 Tables organisation

The `rmsd.py` and `rmsf.py` functions both require, along with others, three mandatory inputs: a filename string, a topology and a trajectory files, which allow the function to generate a unique data table after each call, unless the corresponding file already exists. If this is not the case the function computes the requested metric and outputs the results as a CSV table, labelled as follows:

- *filename_rmsd.csv* for RMSD
- *filename_rmsf.csv* for RMSF.

The *filename* component serves as a unique identifier, encompassing both the system under analysis and the specific structural subset from which the metric is derived. In this work, the filename is systematically constructed from four elements: structure, replica, selection, and segid separated by an underscore character (an example is reported in Table 3). The metric suffixes (`rmsd.csv` and `rmsf.csv`) are automatically appended by the respective functions. Lastly, Table 4 reports the internal organisation of the tables generated by the two metrics.

1. "Structure" refers to the simulated molecule, as defined in Table 2.
2. "Replica" refers to the replica of said structure we are evaluating (in this project each structure has 3 replicas).
3. "Selection", the selection of atoms, or part of the system on which the metric is computed.
4. "Segid", used to identify the subunit we independently evaluated from the others (monomers, ligands and glycans).

An example of this labelling is shown in Table 3.

Structure	Replica	Selection	Segid	Metric	File
apo	1	CA	P0	RMSD	apo_1_CA_P0_rmsd.csv
apo	1	CA	P1	RMSD	...
apo	1	lig	P0	RMSD	
apo	1	lig	P1	RMSD	
apo	1	CA	P0	RMSF	apo_1_CA_P0_rmsf.csv
apo	1	CA	P1	RMSF	...
apo	1	gly	P2	RMSF	
apo	1	gly	P3	RMSF	
apo	1	gly	P4	RMSF	
apo	1	gly	P5	RMSF	
apo	1	gly	P6	RMSF	
apo	1	gly	P7	RMSF	

Table 3: Example of all the tables used in this project and their organisation. To avoid redundancies, here we listed all the tables obtained from a single structure and replica (apo.1), the other protein structures tested here are present in Table 2.

Metric file	Column	Unit	Description
..._rmsd.csv	time	ns	Simulation time
	rmsd	Å	Instantaneous root mean squared displacement
..._rmsf.csv	rmsf	Å	Averaged root mean squared fluctuation
	resid		Residue ID
	segid		Segment ID

Table 4: Interpretation of the columns in the RMSD and RMSF tables.

4.4 RMSD and RMSF evaluation

As shown in the `evaluation.ipynb`, we computed RMSD and RMSF separately on different parts of the system, namely:

- protein, by means of carbon α ;
- glycans;
- ligands, if present.

For each case, we analysed the two copies separately (i.e., monomers, ligand and glycan identical pairs). In all cases, the alignment is carried on the protein structure (using $C\alpha$), ignoring water, ions, ligands and glycan. Before computing the actual metrics the functions perform an alignment to a reference structure of choice, in particular we set:

- in case of RMSD, the structure at frame zero as the reference structure;
- in case of RMSF, the structure obtained by averaging the atomic positions throughout the entire simulation as reference structure.

RMSD is used here to obtain information about the movements of a particular element, for example we used it to define the time of residency of migalastat in the protein. Instead, RMSF is an average measure of local flexibility, useful to track glycans behaviour and to determine possible structural differences in presence of the different mutations.

4.4.1 Protein

First, we computed RMSD of the protein structure to further confirm what seen in section 4.1. RMSD of monomers belonging to the same protein and replica show a similar but not identical behaviour, whereas the differences are increased, as expected, between different replicas. Results are shown in Figure 5.

We computed the RMSF profiles for the protein structures, as illustrated in Figure 6. Calculating RMSF for each monomer independently allowed us to eliminate the contribution of inter-monomer distance fluctuations, and to isolate the intrinsic flexibility of each subunit. Notably, all analyzed structures, including the N215S mutants, exhibit peaks of slightly increased flexibility in correspondence with the glycosylation sites suggesting that the presence or absence of the glycan does not significantly affect the local flexibility of the region. Additional regions of interest include a pronounced peak between residues 52 and 62, which is markedly higher in both the wild-type DGJ and the R301Q mutant, and the segment spanning residues 242 to 262, which displays substantial variability across replicas of the same structure, which, being identical, are expected to behave similarly.



Figure 5: RMSD of protein monomers (computed from carbon α). Monomers belonging to the same replica have the same color.

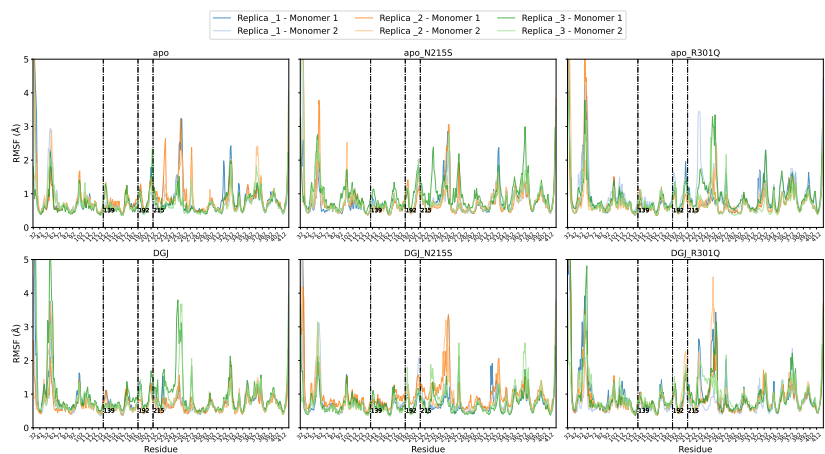


Figure 6: RMSF of protein monomers (computed from α carbons). Monomers belonging to the same replica have the same color.

4.4.2 Glycans

For the glycan analysis we used RMSF to track their mobility in space, as they are attached to the protein by a single covalent bond. As shown in Figure 7, RMSF value tend to increase throughout the glycan structure, as the distance from the N-glycosylation site increases. In some cases the glycan pairs attached to the same residues on the two monomers show more similarity in their behaviour compared to the other structures but it must be considered an artefact given their high mobility.



Figure 7: RMSF of the glycan structures described in Figure 2. Each plot shows three pairs glycan structures (same color), symmetrically distributed between the two monomers, and attached to residues 139, 192 and 215. In case of N215S mutant, the N-glycosylation on that pair of residues is not present. Ticks on the horizontal axis represent the different residues composing the glycan structure (RMSF is computed on atoms).

4.4.3 Migalastat

We were also interested in understanding the ligand behaviour, specifically its interaction with the protein and its residence time (how long it takes to leave its docking site), a key aspect to allow the protein to work correctly once it reaches the lysosomes, as migalastat is a competitor of the true substrate. We used RMSD to track the potential exit of migalastat from the protein, in particular we set at 5 Å the distance at which we consider migalastat out of the binding pocket and the protein. From the initial visualisation of the trajectory in VMD we could already tell that every structure releases at least one of the two ligands approximately towards half of the simulation, which is confirmed by the RMSD plots reported in Figure 8. It is interesting to notice how the R301Q mutant always loses both migalastat and in a shorter time compared to the wild-type and the N215S mutant (for time values, see Table 5).

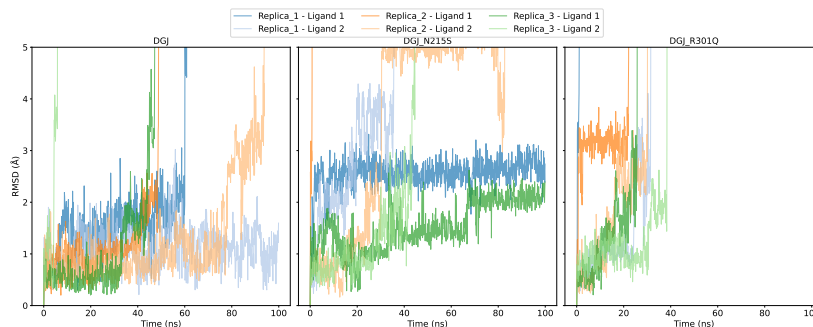


Figure 8: RMSD of individual migalastat. As each monomer contains one ligands, pairs of migalastat from the same replica have the same color.

5 Data Availability

All input structures, simulation protocols, intermediate files, analysis scripts, and processed datasets generated in this study are openly available in the accompanying GitHub repository at github.com/giorginolab/aGAL-MD. The repository includes: (i) system preparation scripts for wild-type and mutant α -galactosidase A, (ii) configuration and run files for MD simulations, (iii) post-processing and analysis notebooks for RMSD, RMSF, and ligand residence time calculations, and (iv) structured data tables and plots derived from the simulations. Additional raw trajectories (in XTC/PSF format) are available on Zenodo at DOI: [10.5281/zenodo.17463313](https://doi.org/10.5281/zenodo.17463313). All materials are released under an open license to facilitate reuse, adaptation, and integration in downstream analyses.

System	Replica	DGJ	Time (ns)
DGJ	1	1	60.1
DGJ	1	2	—
DGJ	2	1	48.9
DGJ	2	2	93.8
DGJ	3	1	47.3
DGJ	3	2	5.9
DGJ_N215S	1	1	35.6
DGJ_N215S	1	2	—
DGJ_N215S	2	1	30.7
DGJ_N215S	2	2	0.9
DGJ_N215S	3	1	44.6
DGJ_N215S	3	2	—
DGJ_R301Q	1	1	31.6
DGJ_R301Q	1	2	1.1
DGJ_R301Q	2	1	30.2
DGJ_R301Q	2	2	22.2
DGJ_R301Q	3	1	38.5
DGJ_R301Q	3	2	25.8

Table 5: Migalastat residence time in each structure, replica and monomer. The exit time is defined as the first time at which migalastat RMSD ≥ 5 Å throughout each simulation. “—” indicates no exit event within 1 μ s.

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References

- [1] Kanako Sugawara et al. “Structural characterization of mutant α -galactosidases causing Fabry disease”. In: *J Hum Genet* 53.9 (Sept. 2008), pp. 812–824. ISSN: 1434-5161, 1435-232X. DOI: 10.1007/s10038-008-0316-9.
- [2] Fumiko Matsuzawa et al. “Fabry disease: correlation between structural changes in α -galactosidase, and clinical and biochemical phenotypes”. In:

Abbreviation	Meaning
α GAL	α -galactosidase A
DGJ	1-Deoxygalactonojirimycin (Migalastat)
MD	Molecular Dynamics
RMSD	Root Mean Square Deviation
RMSF	Root Mean Square Fluctuation
PDB	Protein Data Bank
PDB ID	Protein Data Bank Identifier
WT	Wild Type
HTMD	High-Throughput Molecular Dynamics
NVT	Constant Volume and Temperature Ensemble
NPT	Constant Pressure and Temperature Ensemble
CSV	Comma-Separated Values
PSF	Protein Structure File
DCD	A well-known MD trajectory format
VMD	Visual Molecular Dynamics
N-glycan	N-linked Glycan

Table of abbreviations used throughout the report.

Hum Genet 117.4 (Aug. 2005), pp. 317–328. ISSN: 0340-6717, 1432-1203. DOI: 10.1007/s00439-005-1300-5.

- [3] Michał Nowicki et al. “A review and recommendations for oral chaperone therapy in adult patients with Fabry disease”. In: *Orphanet J Rare Dis* 19.1 (Jan. 2024), p. 16. ISSN: 1750-1172. DOI: 10.1186/s13023-024-03028-w.
- [4] Galafold Amenability, <https://www.galafoldamenabilitytable.com>.
- [5] Raquel L. Lieberman et al. “Effects of pH and Iminosugar Pharmacological Chaperones on Lysosomal Glycosidase Structure and Stability”. In: *Biochemistry* 48.22 (June 2009), pp. 4816–4827. ISSN: 0006-2960, 1520-4995. DOI: 10.1021/bi9002265.
- [6] Erin S. Stokes, M. Lane Gilchrist, and David H. Calhoun. “Prediction of improved therapeutics for fabry disease patients generated by mutagenesis of the α -galactosidase A active site, dimer interface, and glycosylation region”. In: *Protein Expression and Purification* 175 (Nov. 2020), p. 105710. ISSN: 10465928. DOI: 10.1016/j.pep.2020.105710.
- [7] Huang-Yi Li et al. “Mechanistic Insights into Dibasic Iminosugars as pH-Selective Pharmacological Chaperones to Stabilize Human α -Galactosidase”. In: *JACS Au* 4.3 (Mar. 2024), pp. 908–918. ISSN: 2691-3704, 2691-3704. DOI: 10.1021/jacsau.3c00684.
- [8] Daniel Oder et al. “ α -Galactosidase A Genotype N215S Induces a Specific Cardiac Variant of Fabry Disease”. In: *Circ Cardiovasc Genet* 10.5 (Oct.

- 2017), e001691. ISSN: 1942-325X, 1942-3268. DOI: 10.1161/CIRCGENETICS.116.001691.
- [9] Bun Sheng et al. “Two related Chinese Fabry disease patients with a p.N215S pathological variant who presented with nephropathy”. In: *Molecular Genetics and Metabolism Reports* 24 (Sept. 2020), p. 100596. ISSN: 22144269. DOI: 10.1016/j.ymgmr.2020.100596.
 - [10] Mathura Kugan et al. “Fabry disease Enzyme Enhancement on migalastat Study: FEES”. In: *Clinica Chimica Acta* 561 (July 2024), p. 119843. ISSN: 00098981. DOI: 10.1016/j.cca.2024.119843.
 - [11] Jian-Qiang Fan et al. “Accelerated transport and maturation of lysosomal α -galactosidase A in Fabry lymphoblasts by an enzyme inhibitor”. In: *Nat Med* 5.1 (Jan. 1999), pp. 112–115. ISSN: 1078-8956, 1546-170X. DOI: 10.1038/4801.
 - [12] Satoshi Ishii et al. “Mutant α -galactosidase A enzymes identified in Fabry disease patients with residual enzyme activity: biochemical characterization and restoration of normal intracellular processing by 1-deoxygalactonojirimycin”. In: *Biochemical Journal* 406.2 (Sept. 2007), pp. 285–295. ISSN: 0264-6021, 1470-8728. DOI: 10.1042/BJ20070479.
 - [13] M. Brady et al. “Diagnosing Fabry disease—delays and difficulties within discordant siblings”. In: *QJM* 108.7 (July 2015), pp. 585–590. ISSN: 1460-2725, 1460-2393. DOI: 10.1093/qjmed/hct024.
 - [14] Robert B Best et al. “Optimization of the additive CHARMM all-atom protein force field targeting improved sampling of the backbone ϕ , ψ and side-chain χ_1 and χ_2 dihedral angles”. In: *Journal of Chemical Theory and Computation* (2012). ISSN: 1549-9618. DOI: 10.1021/ct300400x.
 - [15] K. Vanommeslaeghe and A. D. MacKerell. “Automation of the CHARMM General Force Field (CGenFF) I: Bond Perception and Atom Typing”. In: *Journal of Chemical Information and Modeling* 52.12 (Dec. 2012), pp. 3144–3154. ISSN: 1549-9596. DOI: 10.1021/ci300363c.
 - [16] Gerard Martínez-Rosell, Toni Giorgino, and Gianni De Fabritiis. “Play-Molecule ProteinPrepare: A Web Application for Protein Preparation for Molecular Dynamics Simulations”. In: *Journal of Chemical Information and Modeling* 57.7 (July 2017), pp. 1511–1516. ISSN: 1549-9596. DOI: 10.1021/acs.jcim.7b00190.
 - [17] William Humphrey, Andrew Dalke, and Klaus Schulten. “VMD: Visual molecular dynamics”. In: *Journal of Molecular Graphics* 14.1 (1996), pp. 33–38. ISSN: 0263-7855. DOI: [https://doi.org/10.1016/0263-7855\(96\)00018-5](https://doi.org/10.1016/0263-7855(96)00018-5).
 - [18] Stefan Doerr et al. “HTMD: High-Throughput Molecular Dynamics for Molecular Discovery”. In: *Journal of Chemical Theory and Computation* 12.4 (2016), pp. 1845–1852. DOI: 10.1021/acs.jctc.6b00049.

- [19] M. J. Harvey, G. Giupponi, and G. De Fabritiis. “ACEMD: Accelerating Biomolecular Dynamics in the Microsecond Time Scale”. In: *Journal of Chemical Theory and Computation* 5.6 (June 2009), pp. 1632–1639. ISSN: 1549-9618. DOI: 10.1021/ct9000685.