

Theoretical Protein Structure Prediction of Glucagon-like Peptide 2 Receptor Using Homology Modelling

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Abstract

Glucagon-like peptide 2 receptor, a GPCR, binds with the glucagon-like peptide, GLP-2 and regulates various metabolic functions in the gastrointestinal tract. It plays an important role in the nutrient homeostasis related to nutrient assimilation by regulating mucosal epithelium. GLP-2 receptor affects the cellular response to external injury, by controlling the intestinal crypt cell proliferation. As they are therapeutically attractive towards diseases related with the gastrointestinal tract, it becomes essential to analyse their structural features to study the pathophysiology of the diseases. As the three dimensional structure of the protein is not available, in this study, we have performed the homology modelling of the receptor based on single- and multiple template modeling. The models were subjected to model validation and a reliable model based on the validation statistics was identified. The predicted model could be useful in studying the structural features of GLP-2 receptor and their role in various diseases related to them.

Keywords: GLP2 Receptor, GPCR, GLP2, Homology Modelling.

1. Introduction

Glucagon and glucagon like peptides (GLPs) are secreted in the pancreas, gut, CNS and PNS and control various metabolic functions^[1]. They are derived from a common precursor, known as proglucagon, and later gives rise to GLPs-1 and -2 in the endocrine cells present in the gastrointestinal tract and glucagon in the pancreatic cells^[2]. They play major role in the control of nutrient assimilation and hepatic glucose production^[3]. GLPs share considerable sequence homology, their aminoacid identity with glucagon ranging from 21% to 48%. GLPs mediate their functions through a family of G-protein coupled receptors called as Glucagon receptor subfamily. Glucagon-like peptide 2 receptor (GLP-2R) is a one of the receptor in the subfamily, which is involved in the intestinal mucosal growth^[4].

GLP-2, a pleiotropic intestinotropic hormone, secreted in the intestine activates GLP-2R. GLP-2 regulates the

nutrient homeostasis proximal to nutrient assimilation by controlling the stasis of mucosal epithelium. The signalling via GLP-2 receptor regulates the intestinal crypt cell proliferation, directly affecting the cellular response to external injury^[5]. In a study by Drucker et al.^[6], GLP-2 was identified to consistently inducing an increase in the bowel weight and villus growth of the jejunum and ileum, which indicates the importance of GLP-2 in small bowel epithelial proliferation. Hence, GLP-2 is therapeutically attractive towards diseases related to the regulation of mucosal health in the gastrointestinal tract. Also, GPCRs have been in the limelight in the recent pharmacological research, and >30% of all the marketed therapeutics are targeted towards them^[7]. GLP-2R, despite its therapeutic importance, does not have a co crystallised structure. This study aims to predict the three dimensional structure of GLP-2 receptor based on theoretical prediction by homology modelling.

Homology modelling serves as a tool in predicting the three-dimensional conformation of a protein, when only the sequence data of the protein is available. Due to the enormous amount of time required to prepare protein for crystallization using experimental process such as protein expression, purification and crystallization, the number of protein structures resolved experimentally lags behind the sequence data available^[8]. Homol-

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ogy modelling can provide as a tool for the experimental procedures in finding the structure of the protein in a rather short time. In this study, we have developed three-dimensional models of GLP-2R based on homology modelling and validated them. The developed models could provide as a tool for further studies on the structural features and binding features of GLP-2R-GLP-2 interaction.

2. Material and Methods

2.1. Template Selection

Amino acid sequence of the human Glucagon-like

peptide 2 receptor (accession No: O95838) was retrieved from the Uniprot database. To identify the suitable templates for modelling the receptor, a protein BLAST^[9] search was performed against the Protein Data Bank^[10]. Based on sequence identity, query coverage and E-value, 2 different templates were selected. The selected templates were – 5VAI and 5NX2. If the level of sequence identity is above 30%, then up to 90% of the polypeptide conformation tends to be modelled well^[11-13]. Both the templates were having sequence identity $\geq 30\%$, i.e., 47% and 46% respectively. As the identities of the templates were above 30% (Table 1), we have per-

Table 1. The query coverage and identity values of the templates

PDB ID	Max Score	Total score	Query Coverage %	E Value	Identity %
5VAI	367	367	70%	6e-122	47%
5NX2	353	353	70%	5e-177	46%



Fig. 1. Alignment between the target (neuromedin U receptor 1) and template (4BWB).

Table 2. Homology modelling validation results after validation using RMSD, Ramachandran plot, ERRAT and QMEAN

Model No	Modeling Platform	Templates Used	Homology Modeling Validation			ERRAT Overall quality factor	QMEAN	
			RMSD	Ramachandran Plot				
				Number of residues in favored region (%)	Number of residues in allowed region (%)			Number of residues in outlier region (%)
1			0.500	83.8	10.5	5.6	67.679	-6.51
2			0.372	82.4	10.3	7.3	57.802	-5.79
3			0.687	80.9	12.3	6.7	55.679	-7.07
4			0.657	83.1	11.1	5.8	51.965	-6.58
5		5VAI	0.624	79.1	12.0	8.9	51.965	-6.66
6			0.312	77.3	13.6	9.1	51.322	-6.59
7			0.775	80.8	10.5	8.7	52.183	-6.13
8			0.520	80.0	12.2	7.8	50.325	-6.96
9			0.554	83.3	9.1	7.6	57.174	-7.00
10			0.483	95.3	3.3	1.5	66.438	-5.20
11			0.333	96.4	2.5	1.1	65.079	-4.45
12			0.463	95.8	3.3	0.9	65.893	-4.77
13			0.515	95.6	3.4	0.9	58.409	-4.39
14		5NX2	0.421	95.6	2.9	1.5	62.844	-4.68
15			0.246	94.6	3.6	1.8	65.604	-5.32
16			0.509	95.8	3.1	1.1	65.227	-4.94
17			0.340	96.9	2.4	0.7	71.005	-4.20
18			0.394	95.5	4.0	0.5	63.303	-4.72
19	EasyModeller		24.527					
			22.317	88.6	7.4	4.0	66.438	-12.24
20			24.638					
			22.418	88.7	8.3	2.9	65.079	-12.07
21			24.727					
			22.588	86.8	9.3	4.0	65.893	-13.21
22			24.287					
			22.332	88.6	8.3	3.1	58.409	-11.95
23		Both	24.582					
			22.332	90.4	6.7	2.9	62.844	-11.99
24			24.305					
			22.336	87.3	8.2	4.5	65.604	-13.03
25			24.612					
			22.331	89.3	8.2	2.5	65.227	-12.59
26			24.864					
			22.484	91.7	6.0	2.4	71.005	-11.16
27			24.102					
			21.826	87.7	8.2	4.2	63.303	-12.89
28			NA	78.8	14.2	7.1	82.936	-10.44
29			NA	71.9	18.3	9.8	72.846	-13.19
30	ITasser	NA	NA	75.9	15.4	8.7	78.125	-10.00
31			NA	72.8	17.6	9.6	90.257	-11.36
32			NA	66.4	22.9	10.7	76.147	-13.55

formed both single template modelling. Query coverage for the templates was greater than 70%. All of the templates retained the seven transmembrane helix regions, which is the characteristic feature of the GPCR proteins.

2.2. Homology Modelling

Using the modelling platforms, EasyModeller 4.0^[14] and ITasser^[15], the three dimensional structures of human Glucagon-like peptide 2 receptor were predicted. Easy-Modeller 4.0 uses MODELLER 9.12^[16] and Python 2.7.1 in the backend. I-TASSER server, an on-line server, is used for protein structure and function predictions. In the recently concluded CASP (Critical Assessment of Techniques for Protein Structure Prediction), I-TASSER was ranked as number one in the server section^[17]. Initially, the predicted models were validated using the RMSD values. Then, using RAMPAGE web server, Ramachandran plots for the models were plotted^[18]. Ramachandran plot provides a way to visualize backbone dihedral angles ψ against ϕ of amino acid residues in protein structure, which identifies the sterically allowed regions for these angles. Later, validation by QMEAN and ERRAT plots were carried out. QMEAN is a comprehensive scoring function for model quality assessment, which determines the compatibility of the predicted model by assessing the local structural quality of transmembrane protein models using statistical potentials^[19]. ERRAT plots are plotted as a function of the position of a sliding 9-residue window. The error function is based on the statistics of non-bonded atom-atom interactions present in the structure^[20].

3. Results and Discussion

3.1. Model Generation

Using EasyModeller, totally 27 models were developed using EasyModeller, based on both single template and comparative modeling. Five best models predicted by the ITasser server were selected. The sequence alignment of the query with the templates was represented in Fig. 1.

3.2. Model Validation

The predicted models were validated using various validation techniques. Root mean square deviation (RMSD) of all the predicted models with their respective tem-



Fig. 2. Model 17 (Yellow) superimposed with the template 5NX2 (Red).

plate was calculated. Ramachandran plot was generated for each model and the number of residues in favourable, allowed and disallowed region was identified. QMEAN scores and ERRAT plots were developed for the models. The statistics of RMS deviation, QMEAN values, Ramachandran plots and ERRAT are represented in the Table 2. Based on the statistics, from the model 17 developed using Easymodeller was found to be the most reliable among the developed models. Also, all the developed models were found to have similar structure. The superimposition of model 17 with the respective template is represented in Fig. 2. RC plot and ERRAT plots of the selected models were represented in Fig. 3 and Fig. 4, respectively.

4. Conclusion

Three dimensional models for human Glucagon-like peptide 2 receptor were developed using both single and multiple template based approach. Model 17 was selected as best, based on the RMS deviation, Ramachandran plot, ERRAT plot and QMEAN values. Based on the results after model validation, it is found that all the generated models are similar and the structures are reliable. These predicted models would be useful in the studying the interaction of GLP-2 with the GLP-2 receptor. Also, these models may serve as a reliable tool for

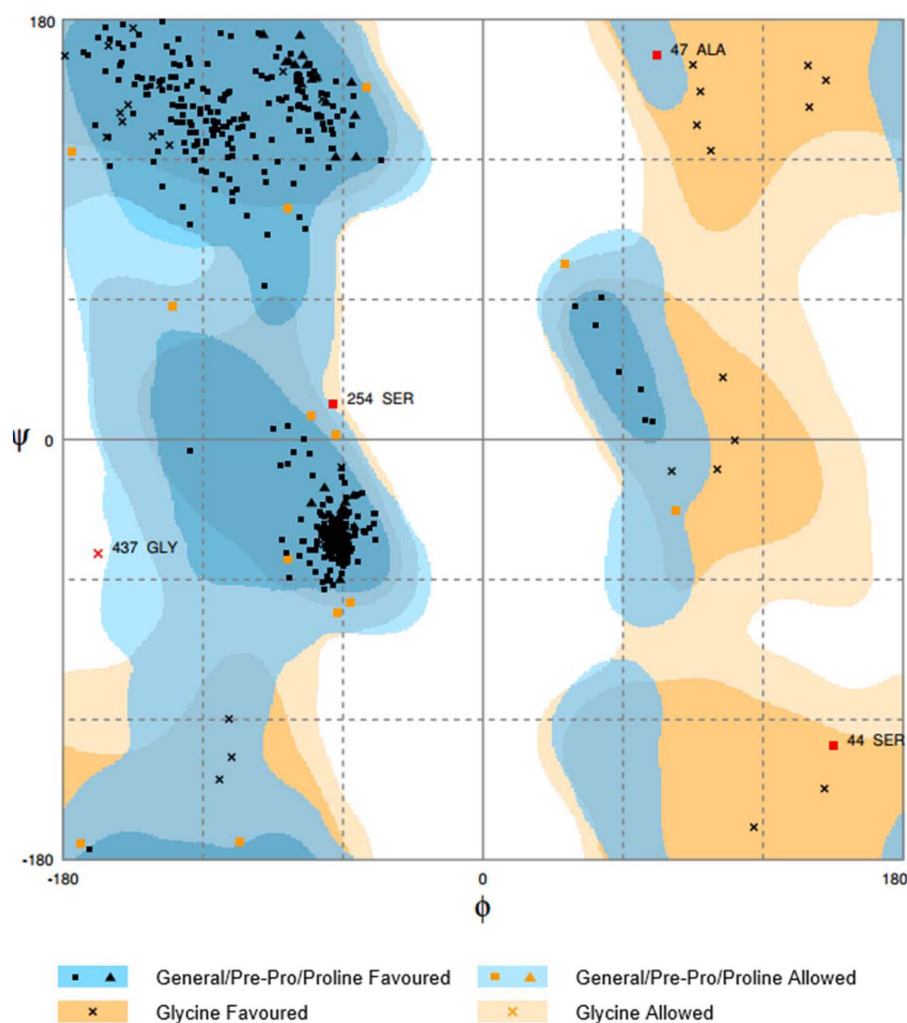


Fig. 3. Ramachandran plot of model 17

Overall quality factor**: 71.005

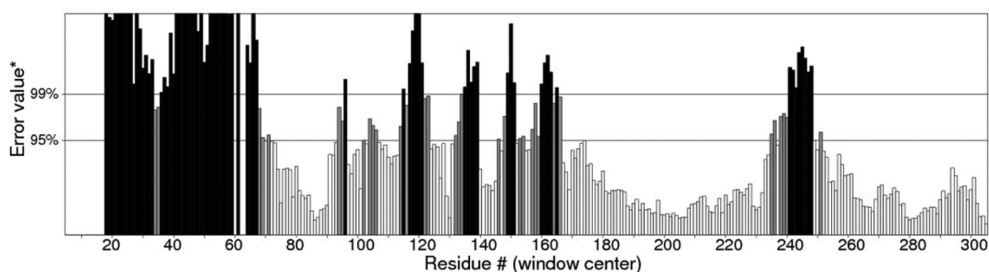


Fig. 4. ERRAT plot developed for the models 17. *on the error axis, two lines are drawn to indicate the confidence with which it is possible to reject regions that exceed that error value, **Expressed as the percentage of the protein for which the calculated error value falls below the 95% rejection limit. Good high resolution structures generally produce values around 95% or higher. For lower resolutions (2.5 to 3 Å) the average overall quality factor is around 91%

analysing the important structural features and function of GLP-2 receptor.

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