Research Article

Molecular Dynamics Simulation and Essential Dynamics of Deleterious Proline 12 Alanine Single-Nucleotide Polymorphism in PPARγ2 Associated with Type 2 Diabetes, Cardiovascular Disease, and Nonalcoholic Fatty Liver Disease

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Background. Peroxisome proliferator-activated receptor-γ (PPARγ) gene is located at 3p25 position. PPARγ functions as the master regulator of glucose homeostasis and lipoprotein metabolism, and recent studies have reported that it is involved in various metabolic diseases such as diabetes mellitus, hyperlipidemia, coronary artery disease (CAD), and nonalcoholic fatty liver disease (NAFLD). PPARγ1 and PPARγ2 are necessary for the development of adipose tissue and insulin sensitivity regulation. But PPARγ2 is the isoform that was controlled in response to nutrient intake and obesity. Methodology. In this study, we used computational techniques to show the impact of Pro12Ala polymorphism on PPARγ2. The 3-D structure of PPARγ2 was modeled using I-TASSER server. The modeled structure was validated with the ZLab server, and the mutation was created with SPDB viewer. Stability prediction tools were used. Molecular dynamics simulation (MDS) was used to understand the structural and functional behavior of the wild type and mutant. Essential dynamics was also applied. Results and Conclusions. Stability prediction tools were showed that this mutation has a destabilizing effect on the PPARγ2 structure. The RMSD, RMSF, Rg, SASA, and DSSP were in line with H-bond results that showed less flexibility in the mutant structure. Essential dynamics was used to verify MDS results. Pro12Ala polymorphism leads to rigidity of the PPARγ2 protein and might disturb the conformational changes and interactions of PPARγ2 and results in type 2 diabetes mellitus (T2DM), CAD, and NAFLD. This study can help scientists to develop a drug therapy against these diseases.

1. Introduction

Peroxisome proliferator-activated receptor-γ (PPARγ) has a key role in adipogenesis, liver and muscle responses to glucose, and pancreatic b-cell function [1]. PPARγ regulates glucose and lipid metabolism. PPARγ has an immune and inflammation suppressive function, which results in an antiatherogenic effect [2, 3]; thus, genetic variation in the PPARγ may regulate individual susceptibility to type 2 diabetes mellitus (T2DM) and coronary heart disease [4]. Alternative splicing of PPARγ results in four isoforms PPARγ1, PPARγ2, PPARγ3, and PPARγ4 in which PPARγ2 is primarily expressed in the adipose tissue [5]. PPARγ2 is a transcription factor that is formed by alternative splicing. Single-nucleotide polymorphisms (SNPs) are widely divided into two distinct clusters, synonymous (sSNPs) and nonsynonymous SNPs (nsSNPs). The nonsynonymous SNPs are further divided into missense mutations and nonsense mutations. The coding synonymous SNPs have a low effect on the protein structure, while the nonsynonymous SNPs have a great impact on the protein structure and higher risk of diseases [6]. The most common single-nucleotide polymorphism PPARγ2 rs1801282 (C>G Pro12Ala) was identified by Yen et al. in 1997 that reduces transcription of
NAFLD – 13 in diabetes mellitus, cirrhosis, liver failure, insulin resistance, and hepato-obesity, hypertension, hypercholesterolemia, type 2 diabetes progressive liver disease that is determined by dyslipidemia, cancer [3]. Nonalcoholic fatty liver disease (NAFLD) is a gene and various diseases such as T2DM, insulin sensitivity, PPAR
mAD database is 0.6 with 1759 being homozygous and fi
Ala12 allele carriers have a significant improvement in insulin sensitivity [2]. Frequency of this mutation in the genomAD database is 0.6 with 1759 being homozygous and 27634 being heterozygous.

Increasing experimental studies have investigated the relationship between Pro12Ala polymorphism of PPARγ2 gene and various diseases such as T2DM, insulin sensitivity, obesity, cardiovascular diseases, Alzheimer’s disease, and cancer [3]. Nonalcoholic fatty liver disease (NAFLD) is a progressive liver disease that is determined by dyslipidemia, obesity, hypertension, hypercholesterolemia, type 2 diabetes mellitus, cirrhosis, liver failure, insulin resistance, and hepatocellular carcinoma [9]. Several researchers reported that Pro12Ala polymorphism was significantly associated with NAFLD [10–13] in different populations.

Coronary artery disease (CAD) is the most common cause of death among diabetic patients. Different studies have also shown which PPARγ2 Pro12Ala polymorphism was associated with CAD [8, 14]. Saremi et al. found PPARγ2 (c.34G>C, Pro12Ala) is considerably associated with higher risk NAFLD [9] in T2DM patients in Iranian population. These findings suggest a high possibility of involvement of Pro12Ala polymorphism in the risk for CAD, NAFLD, and T2DM.

Computational methods were being applied to the study of the structural and functional effect of point mutation at the molecular level. In this investigation, we also implemented multiple computational methods to identify the effect of rs1801282 (C->G Pro12Ala) mutation on PPARγ2 protein. The rs1801282 (C->G Pro12Ala) SNP in the PPARγ2 coding region has an impact on the related disease phenotype, and stability tools were used to survey this polymorphism. The studies have shown that in silico methods can be applied to survey protein structure and function [15–19]. Molecular dynamics simulation (MDS) is an important tool for understanding the effect of mutations on the protein structure, as it provides information about the protein at the atomic level on a reasonable time scale. In order to check (i) whether mutant (12Ala PPARγ2) has an impact on the conformation of the PPARγ2, (ii) whether the mutant structure deviates from the native PPARγ2, and (iii) whether mutant changes flexibility of PPARγ2. We have carried out molecular dynamics simulation of MT (mutant type) and WT (wild type) to predict pathogenic phenotype associated C>G Pro12Ala SNP and further to reveal the conformational flexibility of the mutant PPARγ2 through extensive MDS. At the end, essential dynamics (ED) was applied to the study of the mutant and wild-type proteins. In general, our results provide strong evidence of main conformational drift occurring in Pro12Ala polymorphism as compared to the wild type.

2. Materials and Methods

2.1. Modelling of Protein. rs1801282 (C->G Pro12Ala) SNP was retrieved from dbSNP database (http://www.ncbi.nlm.nih.gov/projects/SNP/, access date: February 19, 2020) [20] for our computational analysis. The amino acid sequence of PPARγ2 with UniProt ID: P37231 was downloaded from UniProt for the study.

We explored RCSB PDB, but there was not a crystallographic structure that included this polymorphism site. Therefore, human PPARγ2 protein (with 505aa) was modeled by an automated protein structure prediction program (I-TASSER) [21]. The modeled structure of PPARγ2 protein was evaluated by the ZLab server (https://zlab.umassmed.edu/bu/rama/index.pl).

We replaced the wild-type protein residue with itself (used as wild-type) and with Alanine (used as mutant) using SPDB viewer [22]. Then, wild type and mutant are minimized with YASARA. In the next step, the effect of this polymorphism on PPARγ2 stability was explored by SNP tools.

2.2. Stability Prediction. Since a missense polymorphism causes alteration of the protein structure and function, therefore, we predicted the protein stability. A number of recent studies have verified implementing multiple bioinformatics tools and algorithms which increase the accuracy of the results [23–29]. To evaluate the effect of the amino acid substitution at position 12 on the stability of wild-type PPARγ2, we used the following stability predictor tools. Mupro is an assembly of programs with machine learning that computes the protein stability and changes based on sequence data, especially when the tertiary structure is not subjected. This approach dominates significant restrictions on previous approaches based on the tertiary structure [30]. DynaMut can perform rapid analysis of the protein stability and dynamics coming from alterations in vibrational entropy [31]. DUET also predicts the effect of point mutations on the protein stability through an embedded computational approach [32]. The mCSM calculates the consequences of missense polymorphisms on the stability of protein, protein-protein binding, and protein-DNA interaction [33]. I-Mutant 2.0 calculations are based on the protein structure or the protein sequence or are based on the prediction of the protein stability of missense variants [34]. The SNAP server was also used. To investigate the mechanism of structural consequences of Pro12Ala mutation on PPARγ2, we performed molecular dynamics simulation.

2.3. Molecular Dynamics Simulation

2.3.1. MD Simulation. To evaluate the deleterious effect of Pro12Ala mutation on the interaction of PPARγ2 protein, we performed molecular dynamics simulation using the actual tool of GROMACS [35]. MDS was carried out with the parallel version of PME in the GROMACS program. The 10 Å nonbonded cut-off was considered. MDS was started with solvation within a dodecahedron-shaped water cage and the 1 nm distance between the cage edges and protein periphery. System neutralization was done with an addition of 13 NA ions. Then, 1000 steps of energy minimization were done. Molecular dynamics simulation was performed at 300 K, 1 atm pressure, and GROMOS53a6 force field using GROMACS 4.6.5 (http://www.gromacs.org/). Before MDS
run, the structure was gained to 300 K the temperature and equilibrated during 100 ps under conditions of constant volume and temperature (NVT). Next, the system was switched to constant pressure and temperature (NPT) and was equilibrated for 100 ps. 50 ns MDS of WT and MT PPARγ2 in 25 × 10^6 steps individually were applied. The cutoff radius of protein-solvent intramolecular hydrogen bonds was 0.35 nm. The periodic boundary condition function was carried out by the leap-frog algorithm with a 2 fs time step and every 500-step structural snapshots (1 ps) [36].

2.3.2. Analysis of Molecular Dynamics Trajectories. Structural deviation analyses of the mutant and wild-type proteins such as root-mean-square deviation (RMSD), root-mean-square fluctuation (RMSF), solvent accessible surface area (SASA), gyration radius (Rg), hydrogen bonds (H-bond), and the secondary structure of the protein (DSSP) were computed using g_rmsd, g_rmsf, g_sasa, g_hbond, g_gyrate, and do_dssp built-in functions of GROMACS package. GRACE software was used for plotting of graphs (http://plasma-gate.weizmann.ac.il/Grace/).

2.4. Essential Dynamics. Essential dynamics, known as Principal Component Analysis (PCA), can show the collective atomic motion of the mutant and wild-type proteins by the GROMACS tool. Principal component analysis was computed using g_covar and g_anaeig built-in functions of the GROMACS package. PCA is a standard protocol for the characterization of eigenvectors and the projection across the first PC1 and PC2 [6].

3. Results

3.1. Modeling SNP Location on Protein Structure. The modeling with I-TASSER gave five models. The best structure with high confidence score was collected and used for further investigations (C-score = -2.38 and estimated TM-score = 0.44 ± 0.14). Model 1 with the highest C-score was selected for further studies. Model 1 was validated by the ZLab server that showed 97.6% of the amino acids of modeled structure in the allowed area (Figure 1), meaning that this model is suitable for further study. The amino acid replacement was also done using SPDB viewer. In the next step, the effect of Pro12Ala polymorphism on the structure and function of PPARγ2 was exhibited by stability prediction tools.

3.2. Stability Prediction. Most of disease-associated polymorphisms have a significant influence on protein stability. To characterize the impact of Pro12Ala SNP on the PPARγ2 structure and function, several computational prediction tools were used. IMutant2.0 has predicted Pro12Ala polymorphism decreases the stability of PPARγ2. mCSM, DUET, and DynaMut also showed the destabilizing impact of Pro12Ala polymorphism on PPARγ2 with DDG (-1.112, -0.6145, and -0.677 cal/mol, respectively). SNAP was also shown, in which this polymorphism is pathogenic with 85% expected accuracy. In the next step, we studied molecular dynamics simulation of the mutant and wild type.

3.3. MD Simulation. Now, computational analysis is a roadmap to define a standard disease-specific SNP at the molecular level. In this study, we examined rs1801282 (C>G Pro12Ala) PPARγ2 which is related to several diseases especially CAD, NAFLD, and T2DM. MDS approaches are also extensively used to report the structural consequences of the deleterious predicted point mutations. We calculated respective Ca-root mean square deviation (Ca-RMSD) for simulations that are plotted in Figure 2(a). WT RMSD plot equilibrated in 0.74 nm while MT RMSD plot equilibrated in 0.65. It represents mutation result in less flexibility of PPARγ2. To understand how mutant affects the dynamic behavior of the residues and to examine the cause of conformational drifts observed in RMSD, Ca-root mean square fluctuation (Ca-RMSF) of WT and MT amino acid residues were calculated and are plotted in Figure 2(b). Lower fluctuations were seen in MT compared to WT. Then, this mutation decreases the flexibility of the PPARγ2 protein.

SASA is a representative of the grade in which an amino acid interacts with its environment (solvent and protein) [37]. An increase or decrease in SASA plot displays changes in the subjected amino acid residues, so affecting the protein tertiary structure. Results of the analysis showed that SASA of WT was 152 nm^2 and SASA of MT was 159 nm^2, which showed a less average total SASA in WT compared to the MT (Figure 3(a)). Rg is also a parameter to explain the equilibrium conformation of a total system specifically in analyzing proteins. It is indicative of the compression level of the protein structure; i.e., polypeptide chain was folded or was unfolded [36]. We observed a more decrease in Rg of MT compared to WT (Figure 3(b)). WT Rg decreased to 2.41 nm while MT Rg decreased to 2.38 nm; then, MT has more compactness and more rigidity structure.
Figure 2: GROMACS analysis of Backbone RMSD and RMSF as a function of time for the mutant and wild-type at 50 ns molecular dynamics simulation; red: mutant; black: wild type; A: RMSD; B: RMSF.

Figure 3: GROMACS analysis of Rg, SASA, and intramolecular hydrogen bonds of Cα atoms for the mutant and wild type in the PPARγ2 protein at 300 K. (a) Rg, (b) SASA, and (c) intramolecular hydrogen bonds. Mutant was shown in black and wild type in red.
One of the main factors that account for maintaining the stable conformation of a protein is hydrogen bonding [36]. We have performed the NH bond analysis of WT and MT during simulations that are plotted in Figure 3(c). Results showed a significant difference in intramolecular hydrogen bond pattern between WT and MT. A greater average number of hydrogen bonds was observed in the MT (340) compared to the WT (340) during simulation, indicating that the occurrence of this mutation may lead to a more compact conformation and rigidity of PPARγ2 protein (Figure 3(c)).

In the end, the secondary structure of the mutant and wild-type proteins was considered that indicated negligible differences (Figure 4). Generally, this mutation decreases flexibility and stability of PPARγ2 which results in the decrease of PPARγ2 expression.

### 3.4. Essential Dynamics Analysis

In this step, we used essential dynamics analysis to obtain the dynamics of the mutant and wild-type proteins. The projection of trajectories of the mutant and wild-type proteins during the essential dynamics in the phase space along the first two principal components (PC1, PC2) at 300 K is plotted in Figure 5. It predicts the large-scale collective motions for the mutant and wild-type of the PPARγ2 protein. PCA analysis showed in which, due to mutation, the structural dynamics is changing. Figure 5(a) plot clearly indicates the mutant occupied less space in phase space while the wild-type occupied more space. The first 50 eigenvectors were selected to compute concerted motions (Figure 5(b)). The eigenvalues were obtained from the diagonalization of the covariance matrix of atomic fluctuations. We observed increased flexibility of the wild type than the mutant; then, Pro12Ala polymorphism causes the rigidity. The PCA analysis results agree with the results from MDS.

### 4. Discussion

Studying mutations can help to comprehend the disease’s molecular mechanism which is associated with their inheri-

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**Figure 4:** Secondary structural elements for the mutant and wild type in the PPARγ2 protein. The color of each secondary structure is displayed in the legend.

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**Figure 5:** (a) The projection of trajectories of the mutant and wild-type proteins during the essential dynamics in the phase space along the first two principal components (PC1, PC2) at 300 K. (b) The first 50 eigenvectors were selected to compute concerted motions.
that may provide insight for therapeutic approaches, especially in T2DM and CAD or NAFLD.

The present study will offer an in-depth insight into the genotype-phenotype association of deleterious SNP rs1801282 (C>G Pro12Ala) in PPARγ2. The flexibility loss is specifically observed in RMSD, RMSF, and Rg plots that showed Ala12 allele has a major impact on the structural conformation of the PPARγ2 protein. A higher number of H-bonds were observed in the PPARγ2 mutant than in the wild-type protein which might lead to a rigid structure of PPARγ2. Previously, we indicated which G482S leads to rigidity and instability of PPARGC1A protein [6]. Kamaraj and Purohit also showed R326H and R356Q resulting in rigidity of tyrosinase-related protein-1 (TYRP1) protein which might disturb the structural conformation and catalytic function of the structure and also play a significant role in inducing Oculocutaneous albinism type III (OCA3) [46]. In the study of Kumar et al. Y63H mutation was also shown with more hydrogen bonds disrupt the wild-type conformation of ATP binding region in CENP-E motor domain [47]. Kumar et al. also suggested that W148R, F161C, and L171R mutations in FLNB might cause the structure to be rigid due to more hydrogen bonds [27].

ED analysis was used for more surveys and was showed a loss of flexibility caused to this polymorphism. Overall, the present computational approach will provide a comprehensive
view of the pathogenic mechanism of rs1801282 SNP, and it may serve as a useful model for predicting the effect of SNPs in other proteins. The results reported in this study elucidate the role of Pro12Ala in PPARγ2 which may provide a useful information for the design of the PPARγ2 mutation-based therapeutic strategies against T2DM.

5. Conclusion

Dysregulation of metabolism is involved in obesity and other diseases like type 2 diabetes mellitus and cardiovascular diseases, which are associated with abnormalities of PPARγ2. PPARγ2 overexpression has been reported to improve type 2 diabetes metabolic and other related conditions [48]. As we mentioned, several researches on PPARγ2 Pro12Ala polymorphism have indicated this polymorphism is pathogenic in different diseases including T2DM, insulin sensitivity, obesity, cardiovascular diseases, Alzheimer’s disease, and cancer [3]. In this study, we also provided evidence of deleterious conformational changes in the PPARγ2 protein that has a significant role in creating disease-associated phenotypes. Ala12 allele represented which leads to disease by changing the structural conformation of PPARγ2.

The stability of a protein is required for its correct function [49–53]. We exhibited a destabilizing effect of this polymorphism using the stability prediction tools. Molecular dynamics simulation was indicated the difference in the dynamics of the PPARγ2 mutant and wild-type proteins. The dynamics of the protein are dependent on the structural flexibility of PPARγ2, and H-bonds are essential to stabilize the protein structure [6]. MD results displayed the decreased flexibility of the Pro12Ala polymorphism structure. The ED analysis also showed this mutation changes the original structural geometry and the structural conformation of PPARγ2 protein, resulting in the loss of the protein function. This suggests that, due to mutation, the structure became more rigid and that this might

Figure 6: Superimposition of secondary structures of mutant and wild-type by UCSF Chimera: (a) whole PPARγ2 protein; (b) mutation site, Pro12 allele (red), and 12Ala allele (blue). Mutation site was highlighted in green.
result in instability and decreasing PPARγ2 expression in patients with related diseases.

The results obtained from this study would facilitate wet-lab researches to develop a potent drug therapy against PPARγ2. The results of this study report the role of Pro12Ala polymorphism in PPARγ2 and may provide useful information for the design of Pro12Ala polymorphism-based therapeutic strategies against especially CAD, NAFLD, T2DM, insulin sensitivity, obesity, cardiovascular disease, Alzheimer's disease, and cancer.

Data Availability

Data is available on request.

Conflicts of Interest

We declare that we have no conflict of interest.

Acknowledgments

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