

# Exploration of Potential Ewing Sarcoma Drugs from FDA-Approved Pharmaceuticals through Computational Drug Repositioning, Pharmacogenomics, Molecular Docking, and MD Simulation Studies

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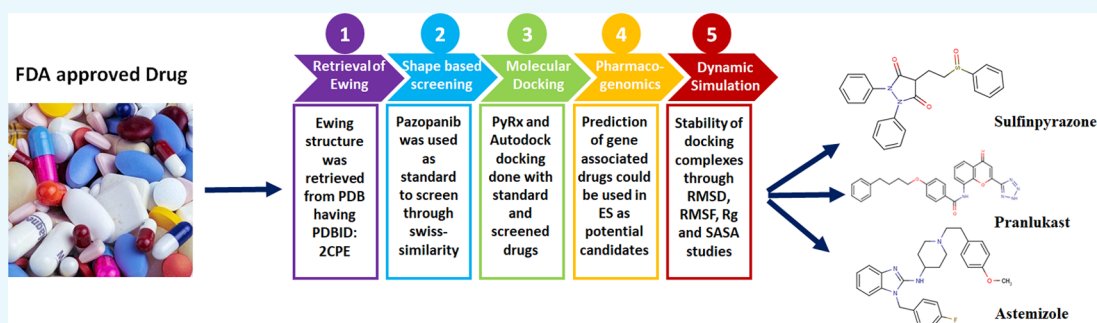
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**ABSTRACT:** Novel drug development is a time-consuming process with relatively high debilitating costs. To overcome this problem, computational drug repositioning approaches are being used to predict the possible therapeutic scaffolds against different diseases. In the current study, computational drug repositioning approaches were employed to fetch the promising drugs from the pool of FDA-approved drugs against Ewing sarcoma. The binding interaction patterns and conformational behaviors of screened drugs within the active region of Ewing sarcoma protein (EWS) were confirmed through molecular docking profiles. Furthermore, pharmacogenomics analysis was employed to check the possible associations of selected drugs with Ewing sarcoma genes. Moreover, the stability behavior of selected docked complexes (drugs-EWS) was checked by molecular dynamics simulations. Taken together, astemizole, sulfapyrazone, and pranlukast exhibited a result comparable to pazopanib and can be used as a possible therapeutic agent in the treatment of Ewing sarcoma.

## 1. INTRODUCTION

Drug development is a time-consuming and overpriced process with particularly low success and relatively high failure rates. To overcome such problems, there are multiple computational drug-designing approaches, including drug repositioning that is being used nowadays.<sup>1</sup> Drug repositioning approaches assist in minimizing the cost and time of the drug development process due to their known efficacy and therapeutic potential against other diseases.<sup>2</sup> There are various computational methods such as feature-based methods, matrix decomposition-based methods, network-based methods, and reverse transcriptome-based methods for drug repositioning.<sup>3,4</sup> However, it has been observed that drug development efforts for the treatment of Ewing sarcoma (ES) have been largely unsuccessful in the last decade.<sup>5</sup>

ES is a cancerous tumor usually observed in bones and other soft tissues like cartilages and nerve tissues, respectively.<sup>6</sup> There are different types of ES based on the position of the tumor within the body, such as extrasosseous and bone sarcomas, skin tumor, and peripheral primitive neuroectodermal tumor (pPNET). ES usually occurs in the pelvic region, shoulder blades, ribs, and femur bones.<sup>7,8</sup> The major symptoms

of ES are long-lasting fever, pain in bones, swelling of muscular and nerve tissues, and stiffness of long bones.<sup>9</sup> It has been observed that Ewing tumors account for 10% of malignancies in humans and metastasize to the other parts of the body more frequently like bone marrow and lungs.<sup>10</sup> A prior research report showed that the Ewing sarcoma protein (EWS) is the basic target of ES and is directly involved in the formation of ES bone carcinogenesis.<sup>11</sup> EWS is an RNA binding protein that binds to Friend leukemia integration 1 transcription factor FLI1 forming EWS/FLI1 fusion protein. The N-terminus of the EWS/FLI1 domain allows EWS/FLI1 to bind with RNA polymerase II and recruit the barrier-to-autointegration factor complex. However, the C-terminus of EWS/FLI1 retains the DNA-binding domain of FLI1 and particularly binds with the

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ACCGGAAG core sequence. The EWS/FLI1 is preferentially bound to GGAA-repetitive regions, and there is a positive correlation between the GGAA microsatellites, EWS/FLI1 binding, and target gene expression.<sup>12</sup>

In the current study, a drug repositioning approach is employed to screen the Food and Drug Administration (FDA)-approved drugs against ES. The human Ewing protein is used as a receptor molecule to screen FDA-approved drugs through a shape-based screening approach. Pazopanib was used as a standard template to access similar ligand structures from FDA-approved compounds through the SwissSimilarity approach. The screened hits having similar chemical structures were accessed from the FDA-approved list and underwent molecular docking analysis using PyRx. Moreover, pharmacogenomics analysis (drug–genes interactions) was carried out by checking all possible genes against all selected drugs. The best-selected drugs were again examined by a docking procedure with AutoDock to check their binding affinities against the Ewing sarcoma protein. Finally, the best-generated docked complexes were further analyzed through molecular dynamics simulations to observe the structural stability through RMSD, RMSF,  $R_g$ , and SASA graphs.

## 2. COMPUTATIONAL METHODOLOGY

**2.1. Retrieval of Protein Structure.** The three-dimensional (3D) solution structure of the RNA recognition motif of the Ewing Sarcoma (EWS) protein having PDBID 2CPE (<https://www.rcsb.org/structure/2CPE>) was retrieved from the Protein Data Bank and its energy was minimized with UCSF Chimera 1.10.1 using conjugate gradient algorithm and Amber force field.<sup>13</sup> The structural assessment of the Ewing sarcoma protein such as  $\alpha$ -helices,  $\beta$ -sheets, coils, and turns was confirmed through the VADAR 1.8 (<http://vadar.wishartlab.com/>) server. The Discovery Studio 2.1.0 Client was used to view the 3D structure of the target protein and for the generation of Ramachandran graphs.<sup>14</sup>

**2.2. Shape-Based Screening of FDA-Approved Drugs Using SwissSimilarity.** The SwissSimilarity<sup>15</sup> is an online platform that allows one to identify similar chemical hits from FDA and other libraries with respect to the reference template structure. Pazopanib (Votrient) is an anticancer FDA-approved drug that was used as the reference template structure against ES.<sup>16,17</sup> The chemical structure of pazopanib was retrieved from the Drug Bank (DB06589) and used as a template molecule to screen FDA-approved drugs. All of the screened drugs were ranked according to their predicted similarity score values (Table S2). The best-screened drugs were sketched in ACD/ChemSketch and further utilized for docking experiments.

**2.3. Prediction of Active Binding Sites of the Ewing Sarcoma Protein.** The Prankweb (<http://prankweb.cz/>) is an online source that explores the probability of amino acids involved in the formation of active binding sites. The binding pocket information was not available in PDB; therefore, active binding site residues of the Ewing sarcoma protein were predicted using Prankweb.

**2.4. Molecular Docking Using PyRx and AutoDock.** Before conducting our docking experiments, all of the screened drugs were sketched in the ACD/ChemSketch tool and accessed in the mol format. Furthermore, the UCSF Chimera 1.10.1 tool was employed for energy minimization of each ligand having default parameters such as steepest descent and conjugate gradient with 100 steps with a step size of 0.02 (Å),

and the update interval was fixed at 10. In the PyRx docking experiment, all screened drugs were docked with the Ewing sarcoma protein using the default procedure.<sup>18</sup> Before docking, the binding pocket of the target protein was confirmed from Prankweb and literature data. In docking experiments, the grid box dimension values were adjusted as center –  $X = -0.8961$ ,  $Y = -1.6716$ , and  $Z = 0.3732$ , whereas size –  $X = 37.8273$ ,  $Y = 36.5416$ , and  $Z = 36.5756$ , respectively, with the default exhaustiveness value = 8. The grid box size was adjusted on binding pocket residues to allow the ligand to move freely in the search space. Furthermore, the generated docked complexes were keenly analyzed to view their binding conformational poses at the active binding site of the Ewing sarcoma protein. Moreover, these docked complexes were evaluated based on the lowest binding energy (kcal/mol) values and binding interaction patterns between ligands and target proteins. The graphical depictions of all of the docked complexes were accomplished with UCSF Chimera 1.10.1 and Discovery Studio (2.1.0).

Furthermore, another docking experiment was employed on best-screened drugs against the Ewing sarcoma protein using the AutoDock 4.2 tool.<sup>19</sup> In brief, for the receptor protein, the polar hydrogen atoms and Kollman charges were assigned. For the ligand, the Gasteiger partial charges were designated, and nonpolar hydrogen atoms were merged. All of the torsion angles for screened drugs were set free to rotate through the docking experiment. A grid map of  $80 \times 80 \times 80 \text{ \AA}^3$  was adjusted on the binding pocket of the Ewing sarcoma protein to generate the grid map and to obtain the best conformational state of docking. A total of 100 runs were adjusted using docking experiments. The Lamarckian genetic algorithm (LGA) and empirical free energy function were applied by taking docking parameters default. All of the docked complexes were further evaluated on the lowest binding energy (kcal/mol) values, and hydrogen and hydrophobic interaction analysis using Discovery Studio (2.1.0) and UCSF Chimera 1.10.1 was performed.

**2.5. Designing of Pharmacogenomics Networks.** To design the pharmacogenomics network model for best-selected drugs, Drug Gene Interaction Databases (DGIdb) (<https://www.dgldb.org/>) and Drug Signatures Database (DSigDB) (<http://dsigdb.tanlab.org/DSigDBv1.0/>) were employed to obtain the possible list of different disease-associated genes. Furthermore, a detailed literature survey was performed against all predicted genes to identify their involvement in ES. Moreover, clumps of different disease-associated genes were sorted based on Ewing sarcoma, and the remaining disease-associated genes were eliminated from the data set.

**2.6. Molecular Dynamics (MD) Simulations.** The best-screened drug-EWS complexes having good energy values were selected to understand the residual backbone flexibility of protein structure; MD simulations were carried out using the Groningen Machine for Chemicals Simulations (GROMACS) 4.5.4 package<sup>20</sup> with GROMOS 96 force field.<sup>21</sup> The protein topology was designed by pdb2gmx command by employing GROMOS 96 force field. For ligand topology, all three drugs were separated from docking complexes and retrieved in the mol format using UCSF Chimera. Furthermore, SwissParam (<https://www.swissparam.ch/>), an online server, was used to generate ligands topologies files. Finally, both generated topologies (protein and ligand coordinates) were merged to run the simulation. Additionally, a simulation box with a minimum distance to any wall of 10 Å (1.0 nm) was generated

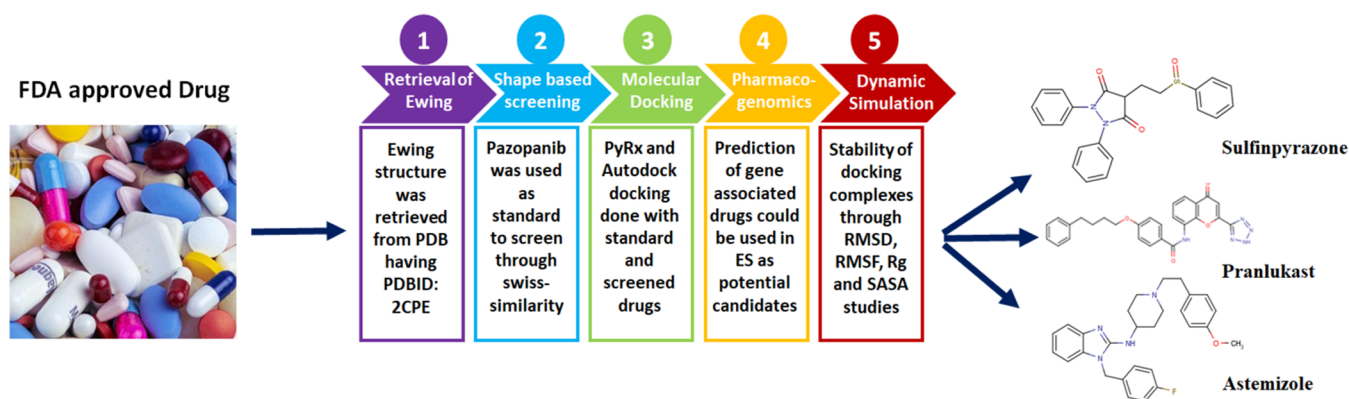


Figure 1. Drug repositioning of ES.

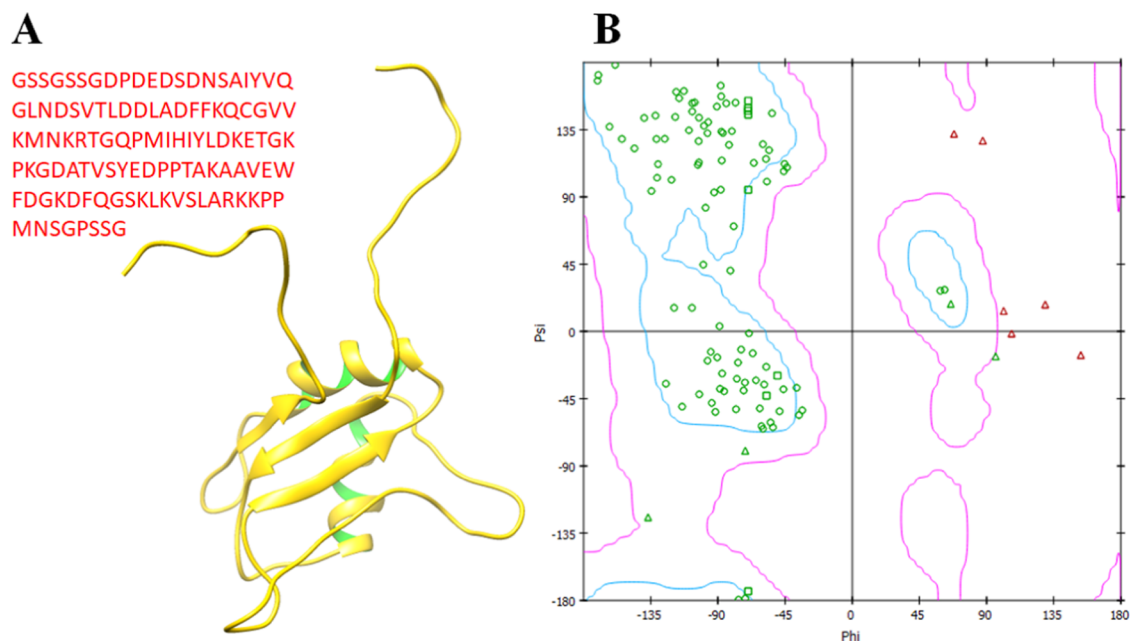


Figure 2. (A, B) 3D structure of the Ewing sarcoma protein with Ramachandran graph.

on the complex by the editconf command. Moreover, the box was filled with solvent molecules using the gmx solvate command by employing the spc216.gro water model. The overall system charge was neutralized by adding ions. The steepest descent approach (1000 ps) for protein structure was applied for energy minimization. For energy minimization, the nsteps = 50 000 were adjusted with an energy step size (emstep) value of 0.01. The Particle Mesh Ewald (PME) method was employed for energy calculation and for electrostatic and van der Waals interactions; cutoff distance for the short-range VdW (rvdw) was set at 14 Å, whereas neighbor list (rlist) and nstlist values were adjusted as 1.0 and 10, respectively, in the em.mdp file.<sup>22</sup> This method permits the use of the Ewald summation at a computational cost comparable to that of a simple truncation method of 10 Å or less, and the linear constraint solver (LINCS)<sup>23</sup> algorithm was used for covalent bond constraints and the time step was set to 0.002 ps. Finally, the molecular dynamics simulation was carried out at 100 ns with nsteps 50 000 000 in the md.mdp file. Different structural evaluations such as root mean square deviations and fluctuations (RMSD/RMSF), solvent accessible surface areas (SASA), and radii of gyration ( $R_g$ ) of backbone

residues were analyzed through Xmgrace software (<http://plasma-gate.weizmann.ac.il/Grace/>) and UCSF Chimera 1.10.1 software.

### 3. RESULTS AND DISCUSSION

The overall design of the research is depicted in Figure 1, showing the flow starting from screening the FDA database into the best-screened drug having good therapeutic potential against ES. Figure 1 shows the different computational evaluation steps such as protein retrieval, drug screening, docking, pharmacogenomic, and MD simulation studies at both protein and drugs level to find out the keen and best possible chemical hits against ES.

**3.1. Structural Assessment of the Ewing Sarcoma Protein.** The Ewing sarcoma protein belongs to a class of hydrolases and consists of a single chain and comprises 113 amino acids (346–458 AA). The overall protein structure shows loops,  $\alpha$ -helices, and  $\beta$ -sheets. It has been observed that two twisted loop structures were present at the terminal regions of the EWS protein and the central binding cavity of helices. Moreover, VADAR 1.8 structure analysis depicted that EWS consists of 25%  $\alpha$ -helices, 30%  $\beta$ -sheets, 43% coils, and



Table 1. SwissSimilarity Scoring Values of FDA-Screened Drugs

drug bank ID	screened drugs	score	drug bank ID	screened drugs	score
DB00310	chlorthalidone	0.005	DB01029	irbesartan	0.014
DB00450	droperidol	0.015	DB01122	ambenonium	0.008
DB00496	darifenacin	0.006	DB01138	sulfinpyrazone	0.007
DB00546	adinazolam	0.007	DB01342	forasartan	0.009
DB00579	mazindol	0.005	DB01349	tasosartan	0.013
DB00637	astemizole	0.006	DB01411	pranlukast	0.007
DB00643	mebendazole	0.006	DB06589	pazopanib	0.998
DB00705	delavirdine	0.010	DB06605	apixaban	0.010
DB00808	indapamide	0.008	DB08828	vismodegib	0.008
DB00837	progabide	0.005	DB08974	flubendazole	0.007
DB00972	azelastine	0.006	DB09003	locapramine	0.006
DB01026	ketoconazole	0.008	DB00280	disopyramide	0.007

20% turns. The Ramachandran plots and values indicated that 93.5% of amino acids exist in the favored region with good accuracy of phi ( $\phi$ ) and psi ( $\psi$ ) angles. Moreover, the coordinates of EWS residues were also plunged into the acceptable region. The overall protein structure and Ramachandran graphs are shown in Figure 2A,B.

**3.2. Shape-Based Screening and Retrieval of Similar Drugs.** In the drug repositioning approach, the shape-based screening, pharmacogenomics and molecular docking simulation are considered significant parameters to predict the possible therapeutic effects of known drugs against different targets.<sup>1,24</sup> Pazopanib was used as a standard drug against EWS<sup>25–27</sup> and used as a template to screen FDA-approved drugs having a similar skeleton. In our computational results, SwissSimilarity results showed 100 FDA-approved drugs that were selected from the pool of 220 FDA-approved drugs based on similarity scoring values ranging from 0.005 to 0.998 (Table S1). Of the 100 FDA-approved drugs, 24 were selected based on best scoring values and have been reported in Table 1. Droperidol (0.015), delavirdine (0.010), irbesartan (0.014), tasosartan (0.013), and apixaban (0.010) showed the highest-scoring values as compared to the rest of all drugs. The screened drugs were ranked based on similarity scoring values, ranging from 0 to 1. The 0 value represents dissimilarity between compounds, whereas 1 is used for highly identical compounds in the screening approach.<sup>1</sup> The chlorthalidone, mazindol, and progabide showed a unique value of similarity score of 0.005 as compared to the standard value. Therefore, 24 drugs were categorized based on the highest, lowest, and medium scoring values and further employed for the docking procedure to check which drug has good binding potential inside the binding pocket of the target protein. Therefore, the selection of drugs was made based on both similarity and docking energy values (Table 1).

Although SwissSimilarity scoring values were low relative to the reference standard value range, structural moieties were similar at different parts in different screened FDA-approved drugs with respect to the standard drug (pazopanib). Therefore, a detailed docking study was run against all screened 24 drugs to check their binding interactions behavior in comparison with pazopanib. Based on these docking results, drugs were selected for further analysis (Figure 3).

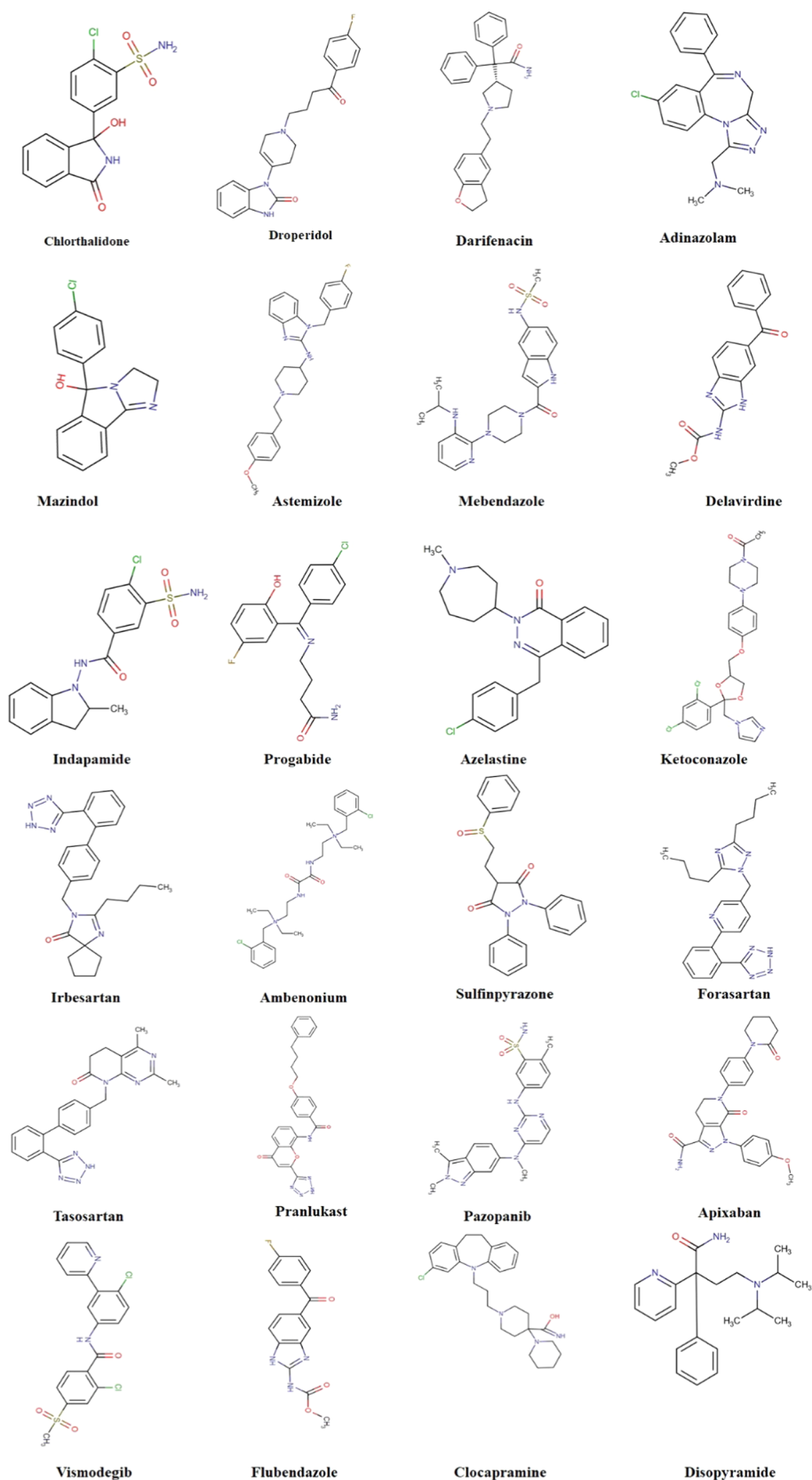
**3.3. Binding Pocket Analysis of the EWS Protein.** The position of a ligand in the holostructure of a protein most probably determines the binding pocket and channels of the target protein.<sup>28</sup> P2Rank is a novel machine learning-based method for the prediction of ligand binding sites inside the

protein structure.<sup>29</sup> PrankWeb, a web server built upon P2Rank, was used by us to explore the binding pockets of the target protein with different pocket sizes and positions inside the target protein. Four different residue binding pockets were predicted based on scoring values (3.52, 2.80, 1.18, and 0.98). The higher pocket score value is 3.52 and constitutes amino acids (Asp359, Asn360, Ser361, Ala362, Lys388, Met397, His399, Tyr401, Thr414, and Ser416) at the central part of EWS. The Soluble Accessible Surface (SAS) area represents the area having a propensity to interact with neighboring atoms. Pocket 1 showed a good SAS value of 50 as compared to other binding pocket values (37, 25, and 21) with different amino acids of EWS (Figure 4). The graphical representation of the binding pocket of EWS is highlighted and depicted in Figure 5A,B.

**3.4. Molecular Docking.** **3.4.1. Binding Affinity Analysis of Screened Drug through PyRx.** Molecular docking is a computational approach used to predict the binding conformational behavior of biomolecules, i.e., drugs and proteins.<sup>30–34</sup> All of the screened drugs were docked and analyzed based on binding affinity (kcal/mol) (see the Supplementary Data S2). From docking results, it has been observed that from 100 FDA-approved drugs, 24 drugs showed binding affinity values higher than  $-7$  kcal/mol and may have good binding potential inside the binding pocket of the EWS protein. The comparative analysis showed that darifenacin (DB00496) exhibited the highest binding affinity value of  $-9.2$ , whereas the rest of the drugs showed values ranging from  $-7$  to  $-9$  kcal/mol (Table 2).

**3.5. Pharmacogenomics Analysis.** **3.5.1. Chlorthalidone, Droperidol, Darifenacin, Adinazolam, and Mazindol.** The best-screened FDA-approved drugs having good binding affinity results were further analyzed through pharmacogenomics analysis. Pharmacogenomics aims to develop rational means to optimize drug therapy with respect to the patients' genotype to ensure maximum efficacy with minimal adverse effects.<sup>35</sup> In our computational analysis, a couple of pharmacogenomics databases were employed to predict the possible links of screened drugs with their respective genes and their association with diseases. The drug's predicted genes were ranked based on interaction score values. In the chlorthalidone–genes network, 10 genes (NPPA, SLC12A1, AGT, SLC12A3, CA1, CA14, CA7, ACE, CA4, and MMP3) were observed with different interaction values and their association with multiple diseases (Table 3). Most of the genes like SLC12A1, AGT, SLC12A3, ACE, and MMP3 are directly involved in the osteosarcoma (a type of cancer that produces





**Figure 3.** Screened FDA-approved drugs.

immature bone). Therefore, chlorthalidone could be used as a repositioned drug against bone cancer by targeting such gene-encoded protein and their associated pathways (Table 3). In droperidol–genes interactions, six genes, DRD2, KCNH2,

DRD4, ADRA1A, DRD3, and CYP2D6, have been observed to have close interaction and involvement in different diseases. The highest predicted interaction of droperidol was observed with DRD2, which is directly linked with osteosarcoma of

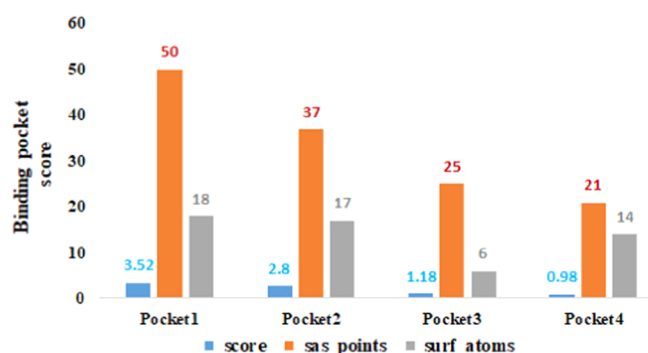


Figure 4. Predicted binding pockets.

bones.<sup>36</sup> Moreover, the predicted HTR2A is also involved in osteosarcoma of bone, especially in childhood.<sup>37</sup> Darifenacin showed interactions with CHRM1/2, CHRM3, CHRM4, CHRMS, and CYP2D6 genes with different correlation values. Darifenacin–CHRM3 showed the highest interaction value (0.61) and was involved in the osteosarcoma of bone.<sup>37</sup>

The Adinazolam–gene network showed that six genes, GABRR1, GABRR3/2, GABRP, GABRE, GABRD, and GABRG1, have a close association with Adinazolam having different interaction values. Literature data showed that these genes are involved in different disorders (Table 3); however, there was no connection between these genes and ES or bone osteosarcoma. Mazindol interacts with SLC6A3, SLC6A2, SLC18A2, SLC6A4, NAT1, and HTT, having good interaction values of 1.27, 0.97, 0.56, 0.21, 0.16, and 0.05, respectively. It was observed that SLC6A2 and SLC18A2 have a close association with childhood osteosarcoma in bone.<sup>38</sup> The osteosarcoma of bone in childhood is the basic characteristic of ES; therefore, our proposed computational research favors the chlorthalidone and mazindol could be used for ES after evaluating and passing through clinical trials.

**3.5.2. Astemizole, Indapamide, Delavirdine, Progabid, Azelastine, and Ketoconazole Pharmacogenomic Analysis.** The astemizole pharmacogenomic analysis showed 10 genes, EED, KCNH1, CYP2J2, HPSE, HRH1, KCNH2, PPAR, ABCB1, CYP3A4, and CYP2D6, having different interaction scoring values (Table 4). Our computational results showed that astemizole has the potential to interact with multiple genes that are directly linked with ES through different ways like mutational or crosslinked signaling pathways. astemizole

showed the highest interaction value (7.57) with EED as compared to other genes that are directly linked to ES.<sup>70</sup> Moreover, astemizole has another interaction with the ABCB1 gene, which possesses a direct role in the etiology of ES. Literature reports also showed that ABCB1 has a good correlation with some diseases such as osteosarcoma of bone, childhood osteosarcoma, sarcoma of soft tissues, fibrosarcoma, adult fibrosarcoma, and peripheral primitive neuroectodermal tumor.<sup>71–75</sup> Another report showed that the KCNH1 and HPSE genes are linked with multiple bone-associated diseases such as childhood osteosarcoma, fibrosarcoma, and synovial sarcoma.<sup>76–80</sup> Indapamide interacts with SLC12A3, KCNE1, KCNQ1, and APEX1 with different scoring values. Prior data showed that SLC12A3 is involved in different diseases such as sarcoma, neoplasms, chondrosarcoma, and adult synovial sarcoma, respectively.<sup>42</sup> Similarly, KCNE1, KCNQ1, and APEX1 are linked with atrial fibrillation,<sup>81</sup> adenocarcinomas,<sup>82</sup> and adenocarcinoma of the lung.<sup>83</sup>

Another screened drug, delavirdine, showed possible interactions with different genes such as ABCG2, ABCC3, ABCC2, ABCC1, and ABCB1 with good interaction scoring values. The literature data reports that all of the genes are involved in the osteosarcoma, osteosarcoma of bone, childhood osteosarcoma, and fibrosarcoma.<sup>84–86</sup> However, among all five genes, ABCC1 is directly involved in ES through different pathways.<sup>87,88</sup> Therefore, our computational results showed that delavirdine could also be used as a good chemical scaffold for the treatment of ES after *in vitro*, *in vivo*, and clinical trials.

Progabid showed interactions with GABBR1 and GABBA1, respectively, which are causative partners of nasopharyngeal carcinoma and osteochondrosis, respectively.<sup>89,90</sup> The drug network showed that azelastine showed interactions with LTC4S, HRH1, HRH2, and PLA2G1B with different interaction values. It has been observed that these genes are myeloid leukemia, chronic atherogenesis, skin carcinoma, and degenerative polyarthritis, respectively. Ketoconazole showed interactions with CYP21A2, CYP3A43, CYP4F2, KCNA10, CYP17A1, ABCG2, NR1I3, CYP3A4, NR1I2, and SNCA. Literature data reported that among all 10 genes, four (ABCG2, NR1I3, CYP3A4, and NR1I2) were involved in childhood osteosarcoma<sup>84,91,92</sup> (Table 4).

**3.5.3. Irbesartan, Amibenonium, Sulfinpyrazone, Forasartan, Tasosartan, Pranlukast, and Gene Interactions.** In

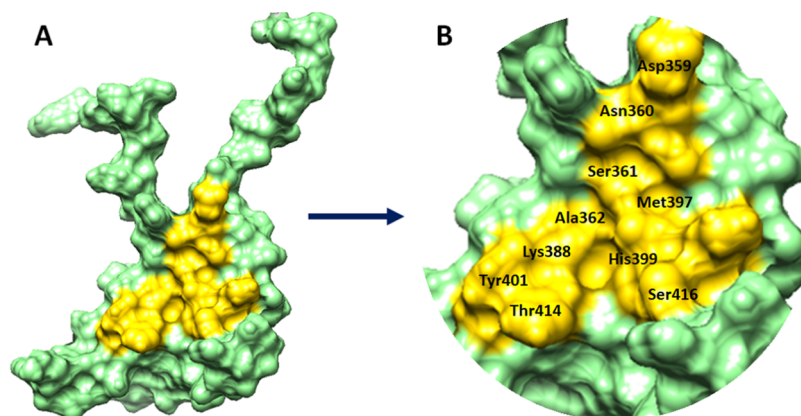


Figure 5. (A, B) Binding pocket of the EWS protein. The EWS protein is represented in cyan color, whereas the binding pocket site is highlighted in yellow color with the labeling of different binding pocket residues.

**Table 2. Binding Affinities of Screened Docking Complexes**

accession numbers	drugs complexes	binding affinity (kcal/mol)	accession numbers	drug complexes	binding affinity (kcal/mol)
DB00310	chlorthalidone	-7.8	DB01029	irbesartan	-7.2
DB00450	droperidol	-7.2	DB01122	ambenonium	-7.2
DB00496	darifenacin	-9.2	DB01138	sulfinpyrazone	-7
DB00546	adinazolam	-7.7	DB01342	forasartan	-7.8
DB00579	mazindol	-7.4	DB01349	tasosartan	-8.9
DB00637	astemizole	-8.3	DB01411	pranlukast	-8
DB00643	mebendazole	-7.4	DB06589	pazopanib	-7.6
DB00705	delavirdine	-7.5	DB06605	apixaban	-7.6
DB00808	indapamide	-7	DB08828	vismodegib	-7.7
DB00837	progabide	-7.2	DB08974	flubendazole	-7.1
DB00972	azelastine	-8.7	DB09003	clocapramine	-8.1
DB01026	ketoconazole	-7.7	DB00280	disopyramide	-7.1

**Table 3. Screened Drugs Chlorthalidone, Droperidol, Darifenacin, Adinozolam, and Mazindol Association with Predicted Genes**

	genes	interaction scores	functions/diseases	references	
chlorthalidone	NPPA	5.68	osteoarthritis, spine	39	
	SLC12A1	3.03	osteosarcoma	40	
	AGT	0.33	childhood osteosarcoma	41	
	SLC12A3	0.32	sarcoma, neoplasms	42	
	CA1	0.44	neoplasms	43	
	CA14	0.41	malignant neoplasms	44	
	CA7	0.36	colorectal carcinoma	45	
	ACE	0.16	synovial sarcoma	46	
	CA4	0.25	retinitis pigmentosa 17	47	
	MMP3	0.21	osteosarcoma of bone; primary osteosarcoma	48, 49	
	droperidol	DRD2	0.35	malignant bone neoplasm; osteosarcoma of bone	36
		KCNH2	0.04	malignant neoplasm of prostate	50
		DRD4	0.11	carcinoma of the lung	51
		ADRA1A	0.09	osteoporosis	52
DRD3		0.08	neoplasms	53	
CYP2D6		0.01	bone cysts, aneurysmal	54	
HTR2A		0.04	osteosarcoma of bone; childhood osteosarcoma	37	
darifenacin		CHRM3	0.61	osteosarcoma of bone	37
	CHRM1/2	0.32	mental depression	55	
	CHRM4	0.46	schizophrenia	56	
	CHRM5	0.45	systemic sclerosis	57	
	CYP2D6	0.03	eosinophilia-myalgia syndrome	58	
	adinazolam	GABRR3/2	0.46	restless legs syndrome	59
GABRR1		0.43	migraine disorders	60	
GABRP		0.22	tumor progression	61	
GABRE		0.22	malignant neoplasms	62	
GABRD		0.21	Rett syndrome	63	
GABRG1		0.21	body height	64	
mazindol		SLC6A3	1.27	scoliosis	65
	SLC6A2	0.97	childhood osteosarcoma; osteosarcoma of bone	66	
	SLC18A2	0.56	childhood osteosarcoma; osteosarcoma of bone	66	
	SLC6A4	0.21	synovial sarcoma	67	
	NAT1	0.16	neoplasms	68	
	HTT	0.05	Huntington Disease	69	

Irbesartan pharmacogenomic analysis, 10 genes (AGTR1, SLC10A1, AGT, EDN1, APOB, APOE, ACE, JUN, SLC2A4, and CYP2C9) were involved in interactions with different scoring values. Among them, AGTR1, AGT, and SLC2A4 are involved in osteosarcoma of bone, childhood osteosarcoma, and osteoarthritis, respectively. However, the rest of the genes were associated with different diseases (Table 5). Ambenonium is associated with the acetylcholinesterase (AChE) gene, which is directly involved in Alzheimer's disease.<sup>113</sup>

Sulfinpyrazone showed interactions with SLC22A12, ABCC1 and ABCC2, FPR1, UGT1A9, NRI12, CYP3A4, VDR, and HPGD, which are involved in renal hypouricemia, fibrosarcoma, carcinogenesis, and different osteosarcomas. Most importantly, the ABCC1 gene is also directly involved in ES in different mechanistic pathways.<sup>87,88</sup> Therefore, computational prediction of the literature data justify that sulfinpyrazone could be used as a screened drug against ES by targeting ABCC1-encoded protein and their associated down-



Table 4. Astemizole, Indapamide, Delavirdine, Progabid, Azelastine, and Ketoconazole Gene Interactions

	genes	interaction scores	functions/diseases	references
astemizole	EED	7.57	Ewing sarcoma	70
	KCNH1	0.95	osteosarcoma of bone; childhood osteosarcoma	76, 77
	CYP2J2	0.76	carcinogenesis	93
	HPSE	0.58	childhood osteosarcoma; fibrosarcoma; synovial sarcoma	78–80
	HRH1	0.27	atherogenesis	94
	KCNH2	0.12	malignant neoplasm of the prostate	50
	PPARD	0.03	obesity	95
	ABCB1	0.01	Ewing's sarcoma; osteosarcoma of bone; childhood osteosarcoma; sarcoma of soft tissue; fibrosarcoma; adult fibrosarcoma; peripheral primitive neuroectodermal tumor	71–75
	CYP3A4	0.01	osteosarcoma	96
	CYP2D6	0.01	bone cysts, aneurysmal	54
indapamide	SLC12A3	3.94	sarcoma, neoplasms, chondrosarcoma, adult synovial sarcoma	42
	KCNE1	7.1	atrial fibrillation	81
	KCNQ1	5.01	adenocarcinoma	82
	APEX1	0.08	adenocarcinoma of the lung	83
delavirdine	ABCG2	0.36	childhood osteosarcoma; osteosarcoma of bone	84
	ABCC3	0.91	carcinoma	85
	ABCC2	0.42	sarcoma, fibrosarcoma	86
	ABCC1	0.42	Ewing's sarcoma of bone; osteosarcoma of bone; childhood osteosarcoma	87, 88
	ABCB1	0.09	osteosarcoma of bone; childhood osteosarcoma	75, 97, 98
progabid	GABBR1	15.49	nasopharyngeal carcinoma	89
	GABRA1	1.76	osteocondrosis	90
azelastine	LTC4S	7.1	myeloid leukemia, chronic	99
	HRH1	1.08	atherogenesis	94
	HRH2	1.58	skin carcinoma	100
	PLA2G1B	0.92	degenerative polyarthritis	101
ketoconazole	CYP21A2	4.37	congenital adrenal hyperplasia	102
	CYP3A43	0.62	carcinogenesis	103
	CYP4F2	0.55	hypertensive disease	104
	KCNA10	0.31	brain neoplasms	105
	CYP17A1	0.31	osteoporosis; carcinoma; spondylarthritis; spondylarthropathies	106–108
	ABCG2	0.07	childhood osteosarcoma; osteosarcoma of bone	84
	NR1I3	0.16	childhood osteosarcoma; osteosarcoma of bone; synovial sarcoma; exostoses	91, 92
	CYP3A4	0.02	childhood osteosarcoma; osteoporosis; osteosarcoma of bone	96, 109
	NR1I2	0.04	osteosarcoma of bone; adolescent idiopathic scoliosis	110, 111
	SNCA	0.07	Parkinson disease 1	112

stream signaling pathways. Furthermore, a couple of other drugs, forasartan and tasosartan, showed interactions with AGTR1 and AGTR2, which are also connected with osteosarcoma of bones.<sup>114,115</sup>

Pranlukast formed a complex with seven different genes such as RNASE3, CYSLTR1, IL5, MUC2, CYSLTR2, TNF, and NFKB1, which are involved in different diseases. Pranlukast–RNASE3 showed the highest interaction value (20.29) as compared to other pharmacogenomics complexes. Moreover, the literature data showed that RNASE3 is involved in ES through different mechanistic pathways. Therefore, computational prediction and literature mining suggest that pranlukast could also be a good therapeutic agent against ES by targeting RNASE3.<sup>116–118</sup>

**3.5.4. Pazopanib, Apixaban, Vismodegib, Clozapramine, and Disopyramide Pharmacogenomic Analysis.** Pazopanib showed interactions with SH2B3, FGF1, ABCG2, and HLA-B, which play a key role in different diseases including leukemia, osteosarcoma of bone, and ankylosing spondylitis, respectively. Apixaban formed a pharmacogenomic complex with three genes such as ABCG2, F10, and CYP3A5, respectively. Literature data showed that these genes are mainly involved in the osteosarcoma of bone in childhood and osteopo-

rosis.<sup>84,139</sup> Vismodegib showed interactions with PTCH1, SMO, and SHH with different interaction scoring values. Prior data reported that these genes are involved in rhabdomyosarcoma and osteoarthritis.<sup>140,141</sup> Moreover, a couple of other screened drugs, clozapramine and disopyramide, also formed pharmacogenomic complexes with different genes, which are involved in different diseases (Table 6).

Based on pharmacogenomics analysis and extensive data mining of five screened FDA-approved drugs, chlorthalidone, astemizole, ketoconazole, sulfapyrazone, and pranlukast were selected for further molecular docking and MD simulation analysis. Figure 6 shows that these five drugs have direct involvement in ES and related bone cancers. Chlorthalidone has genomic interactions with different genes, which are associated with different diseases. Similarly, astemizole has pharmacogenomic interactions with different genes and is associated with long bone cancer (ES), lung cancer, liver, breast, and stomach sarcoma. It has been observed that astemizole may be used as a good therapeutic potential against Ewing sarcoma. The predicted pharmacogenomic results of ketoconazole also showed its therapeutic potential against osteoporosis (bone cancer) and different sarcomas. Sulfapyrazone also has an association with osteoporosis and lung and

**Table 5. Screened Drugs Irbesartan, Ambenonium, Sulfinpyrazone, Forasartan, Tasosartan, Pranlukast and Gene Interactions**

	genes	interaction scores	functions/diseases	references	
irbesartan	AGTR1	1.76	osteosarcoma of bone; childhood osteosarcoma; osteoarthritis	114, 115	
	SLC10A1	4.73	hepatitis B	119	
	AGT	1.11	childhood osteosarcoma	41	
	EDN1	0.79	carcinogenesis	120	
	APOB	0.5	carcinogenesis	121	
	APOE	0.47	Alzheimer's Disease	122	
	ACE	0.2	osteoporosis	123	
	JUN	0.3	osteosarcoma	124	
	SLC2A4	0.22	osteosarcoma of bone; childhood osteosarcoma	125	
	CYP2C9	0.04	ankylosing spondylitis	126	
	ambenonium	ACHE	9.19	Alzheimer's disease	113
		sulfinpyrazone	SLC22A12	2.37	renal hypouricemia
	ABCC1		0.88	Ewing's sarcoma of bone osteosarcoma of bone; childhood osteosarcoma	87, 88
ABCC2	0.79		sarcoma, fibrosarcoma	86	
FPR1	0.59		carcinogenesis	128	
UGT1A9	0.5		carcinoma	129	
NR1I2	0.1		osteosarcoma of bone; adolescent idiopathic scoliosis	110, 111	
CYP3A4	0.02		osteosarcoma	96	
VDR	0.01		osteoporosis	130	
HPGD	0.02		carcinogenesis	131	
forasartan	AGTR1		8.91	osteosarcoma of bone; childhood osteosarcoma; osteoarthritis, knee	114, 115
tasosartan	AGTR1		5.01	osteosarcoma of bone; childhood osteosarcoma; osteoarthritis, knee	114, 115
	AGTR2		7.1	hypertensive disease	132
pranlukast	RNASE3		20.29	Ewing's sarcoma of bone; osteosarcoma of bone; deformity of bone	116–118
	CYSLTR1	6.09	carcinogenesis; adenocarcinoma	133	
	IL5	6.09	asthma	134	
	MUC2	6.09	carcinoma, signet ring cell	135	
	CYSLTR2	1.01	carcinogenesis	136	
	TNF	0.37	rheumatoid arthritis	137	
	NFKB1	0.25	osteosarcoma	138	

**Table 6. Screened Drugs Apixaban, Vismodegib, Clozapramine, Disopyramide, and Genomic Interactions**

	genes	interaction scores	functions/diseases	references
pazopanib	SH2B3	0.84	precursor cell lymphoblastic leukemia lymphoma	142
	FGF1	0.36	osteosarcoma of bone; bone neoplasms	143, 144
	ABCG2	0.04	childhood osteosarcoma of bone	84
	HLA-B	0.15	ankylosing spondylitis	145
apixaban	ABCG2	0.89	childhood osteosarcoma of bone	84
	F10	2.41	osteoporosis	139
	CYP3A5	0.54	neoplasms	146
vismodegib	PTCH1	31.56	rhabdomyosarcoma	140
	SMO	25.25	osteoarthritis of the hip	141
	SHH	12.62	polydactyly	147
clozapramine	PPARD	0.39	obesity	95
disopyramide	KCNH2	0.1	malignant neoplasm of the prostate	50
	CHRM2	0.16	mental depression	55
	CHRM1	0.12	mental depression	55
	CHRM3	0.09	osteosarcoma of bone	37
	CYP2D6	0.01	bone cysts, aneurysmal	54

stomach cancer. The comparative result showed that a couple of screened FDA-approved drugs have a direct association with bone cancer such as ES, which could be further used to check their efficacy through cellular and clinical evaluations.

### 3.6. Binding Affinity Evaluations Using AutoDock.

Based on PyRx docking energy (kcal/mol) and pharmacogenomics analysis, the best five drugs chlorthalidone (DB00310), astemizole (DB00637), ketoconazole (DB01026), sulfinpyrazone (DB01138), and pranlukast (DB01411) have been selected for further binding conformational analysis. The

generated AutoDock results showed the binding energy, drug efficiency, and internal, electrostatic, and torsional energy values (kcal/mol) of best-selected screened drugs (Table 7).

**3.6.1. Superimposition of Screened Drugs within the Active Region of the EWS Protein.** All of the docked structures were superimposed to check the binding configurations of all five screened drugs within the active region of the EWS protein. The binding pocket analysis showed that all of the screened FDA-approved drugs were narrowed in the binding pocket and bound with similar residues with little

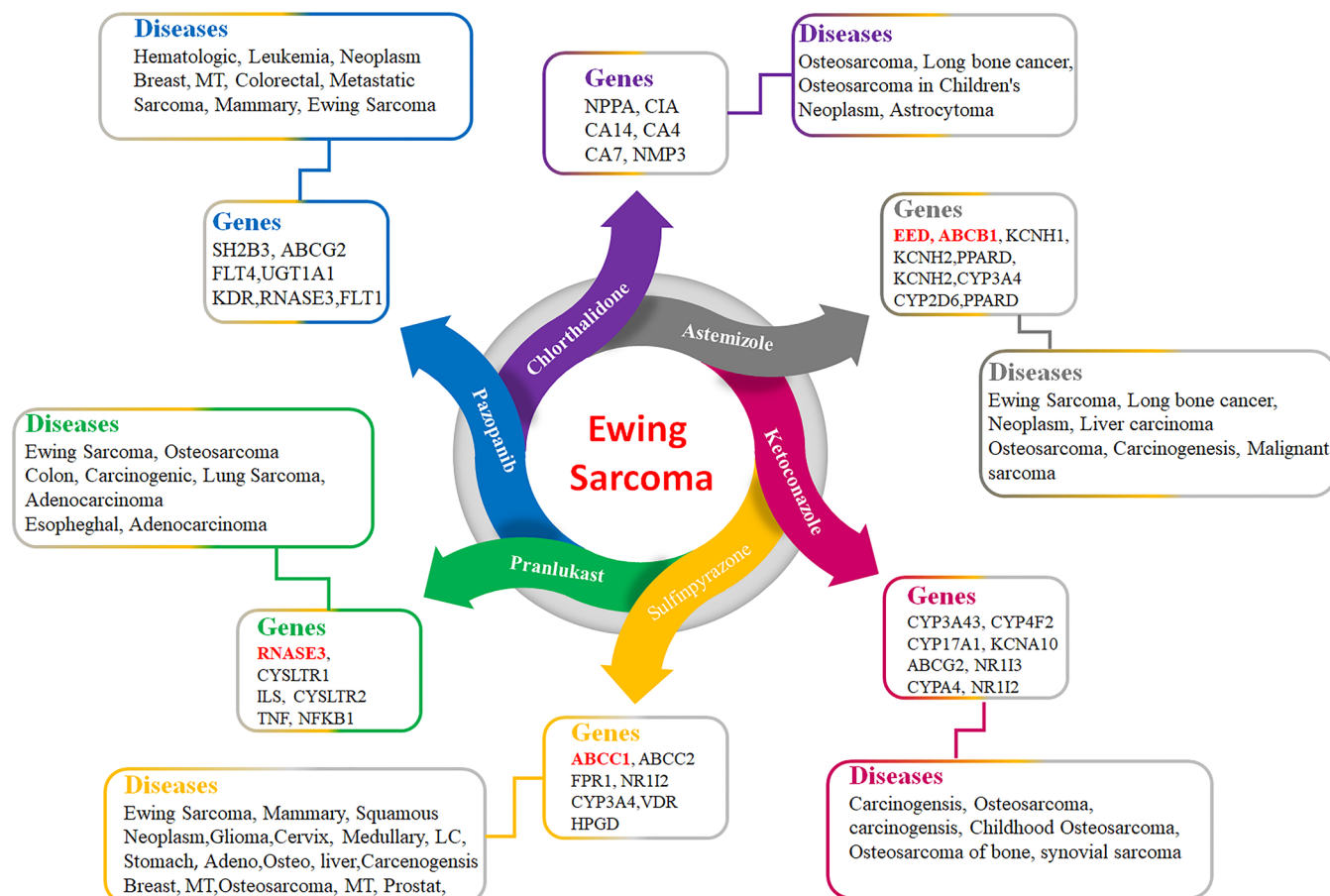


Figure 6. FDA-approved drugs and associations with ES.

Table 7. Binding Affinities of Screened Docking Complexes

drugs	binding energy (kcal/mol)	drug efficacy	internal energy (kcal/mol)	electrostatic (kcal/mol)	torsional energy (kcal/mol)
chlorthalidone	-4.88	0.22	5.18	0.01	0.3
astemizole	-1.20	0.04	3.59	0.01	2.39
ketoconazole	-0.66	0.02	2.15	0.15	1.49
sulfimpyrazone	-5.18	0.18	3.39	0.13	1.79
pranlukast	-11.84	0.33	9.75	0.01	2.09

different conformational poses within the binding pocket of the EWS protein. The binding of all FDA-approved drugs at the same position justified the docking reliability and the accuracy of predicted interactive results (Figure 7).

**3.6.2. Chlorthalidone Hydrogen Binding Analysis.** The chlorthalidone–EWS docked complex is analyzed based on the interaction pattern of binding pocket residues of EWS. The chlorthalidone binds with EWS having a good conformational position inside the active region encompassed by His399, Ile398, Leu374, Thr373, Val372, Pro409, and Leu402 residues, respectively. The oxygen atom of the thiol group in chlorthalidone formed hydrophobic interactions with His399 with a bond distance of 4.10 Å. Moreover, another oxygen atom is present in the five-member ring of the drug forming another hydrogen bond with Thr373 with a bond distance of 2.66 Å. In both chlorthalidone docking interactions both bonds provided good stable behavior to the docking complex, and bond distances were comparable to standard values (<5 hydrophobic and <3 Å hydrogen bonds), respectively (Figure 8).

**3.6.3. Astemizole Hydrogen Binding Analysis.** In astemizole–EWS docking, astemizole binds within the target site of EWS with an appropriate conformational position through interaction with different residues, His399, Ser416, Met397, Thr393, Gln395, Leu374, Ile400, and Ile398. The nitrogen atom of the amino group attached to a heterocyclic group formed a couple of hydrogen bonds with Ile398 and His399, with bond distances of 2.14 and 3.06 Å, respectively. Both astemizole interactions have good comparable values with standard values (<5 hydrophobic and <3 Å hydrogen bonds), respectively (Figure 9).

**3.6.4. Ketoconazole Hydrogen Binding Analysis.** In ketoconazole–EWS docking, the drug binds with site-specific residues with appropriate conformational behavior. Ketoconazole encompassed different Arg392, Asn390, Gln395, Leu374, Thr373, Ile402, Ile400, His399, Ile398, and Met397 residues. Four hydrogen bonds were observed between the ketoconazole–EWS docking complex. The oxygen atom of the benzene ring formed tetrahydrogen bonds at Arg392 and Asn390 with bond distances of 2.04, 2.81, 2.66, and 2.72 Å, respectively.



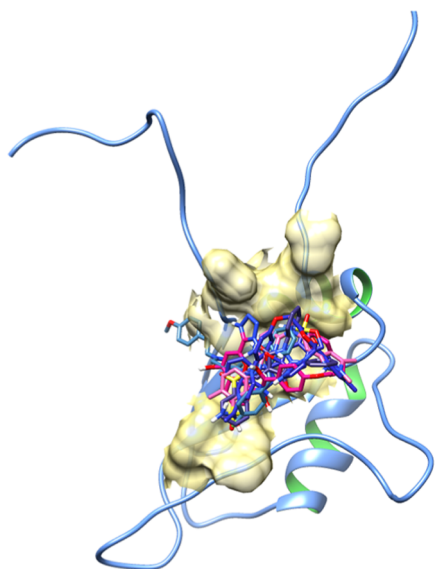


Figure 7. Superimposition of five docking complexes.

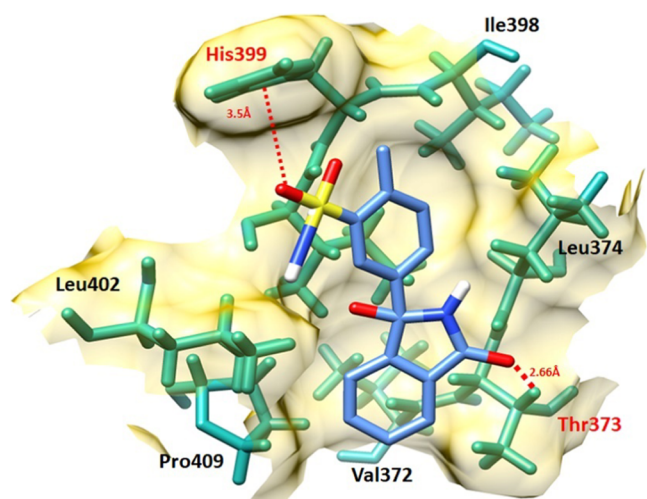


Figure 8. Chlorthalidone–EWS docking complex. The protein structure is represented in gray and purple color, whereas the binding pocket of the EWS protein is highlighted in yellow color in the surface format. The residues are highlighted in dark green color, whereas chlorthalidone is highlighted in blue color and embedded moieties such as oxygen, sulfur, and hydrogen are represented by red, yellow, and light gray colors, respectively.

Ketoconazole interactions with EWS exhibited stable behavior in the docking complex (Figure 10).

**3.6.5. Sulfapyrazone and Pramlukast Hydrogen Binding Analysis.** In sulfapyrazone–EWS and pramlukast–EWS docking complexes, drugs bind with the binding pocket of EWS at slightly deviant conformational positions. The oxygen atom of sulfapyrazone forms a single hydrogen bond with Met397 with a bonding distance of 1.86 Å. However, pramlukast forms two hydrogen bonds with EWS at Met397 and Tyr401 with bonding distances of 2.49 and 1.97 Å, respectively (Figures 11 and 12).

**3.6.6. Pazopanib Docking Analysis.** To check the accuracy of our docking results of screened FDA-approved drugs, the pazopanib–EWS docking complex was analyzed and checked for the interactive behavior against the target protein. The pazopanib-docking behavior showed similar amino acids such as

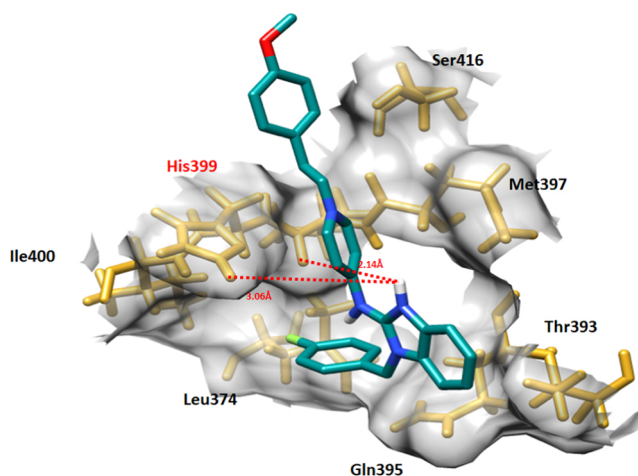


Figure 9. Astemizole–EWS docking complex. The EWS structure is represented in light pink color, whereas the binding pocket of the EWS protein is highlighted in gray color in the surface format. The residues are highlighted in golden color, whereas astemizole is highlighted in blue color and embedded moieties different colors, respectively.

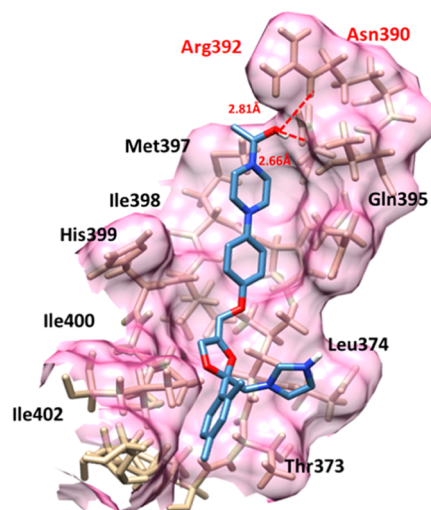
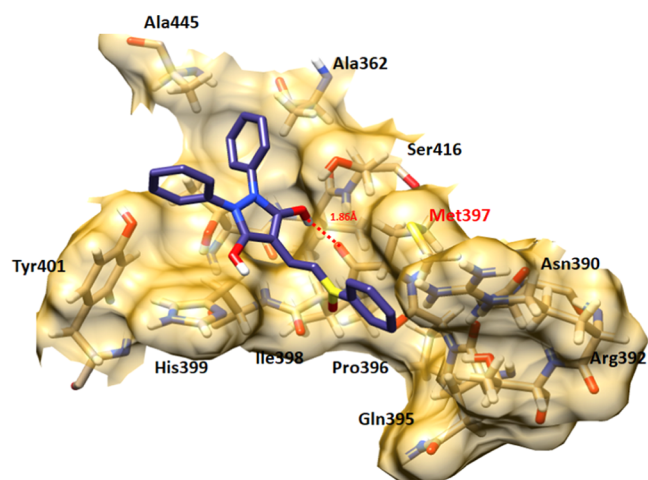


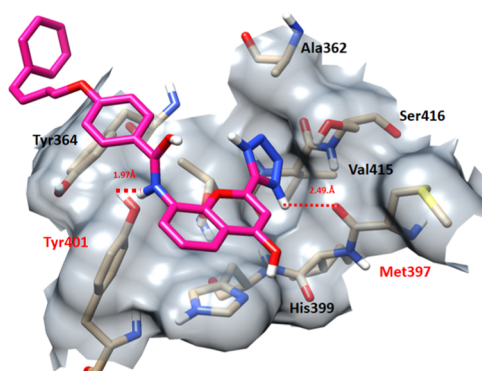
Figure 10. Ketoconazole–EWS docking complex. The interaction residues are highlighted in light pink color, whereas red dotted lines represent the hydrogen bonds in angstrom (Å).

Leu444, Ala445, Ala362, Ser416, Thr414, Tyr401, Tyr364, Gln366, and Ser443. Two hydrogen bonds were observed between the nitrogen of amino ( $\text{NH}_2$ ) and hydrogen atom of the methyl group ( $\text{CH}_3$ ) with Tyr401 and Ala445 with bond distances of 3.13 and 2.16 Å, respectively (Figure 13). The comparative analysis showed that the best five screened drugs bind with the EWS protein in a conformational pattern similar to pazopanib–EWS interactions. Therefore, the screened drugs may be used as a therapeutic template for the designing of novel drugs for the treatment of ES.

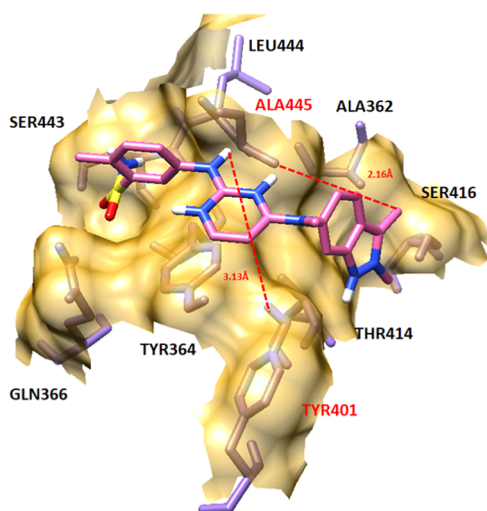
**3.7. Screened Drugs and Their Possible Repositioned Functions.** Based on pharmacogenomics, molecular docking, and detailed literature mining, the selected five drugs, chlorthalidone, astemizole, ketoconazole, sulfapyrazone, and pramlukast, were keenly observed, and their repositioned functions were proposed by targeting different genes. It has been observed that chlorthalidone is usually used in hypertension; however, their proposed repositioned function is to be



**Figure 11.** Sulfapyrazone–EWS docking complex. The interaction residues are highlighted in light pink color, whereas red dotted lines represent the hydrogen bonds in angstrom (Å).



**Figure 12.** Pranlukast–EWS docking complex. The interaction residues are highlighted in light pink color, whereas red dotted lines represent the hydrogen bonds in angstrom (Å).



**Figure 13.** Pazopanib–EWS docking complex.

used against different osteosarcomas of bones and tissues. Moreover, astemizole is frequently used as an antihistamine and prescribed by medical staff against allergies. However, based on pharmacogenomic study and detailed literature mining, it could be used as a new therapeutic agent against ES

by targeting the EED gene. Similarly, ketoconazole is used against Seborrheic dermatitis; however, its proposed repositioned function is osteosarcoma of bone in children. Sulfapyrazone and pranlukast usually are used against Gouty arthritis and allergic rhinitis and asthma, respectively. However, our computational results showed their significance against ES by targeting ABCC1 and RNASE3 genes, respectively. The comparative results showed that astemizole, sulfapyrazone, and pranlukast could be used as new drugs against ES by encompassing animal and clinical approaches (Table 8).

**Table 8. Selected Drugs and Their Involvement in Diseases**

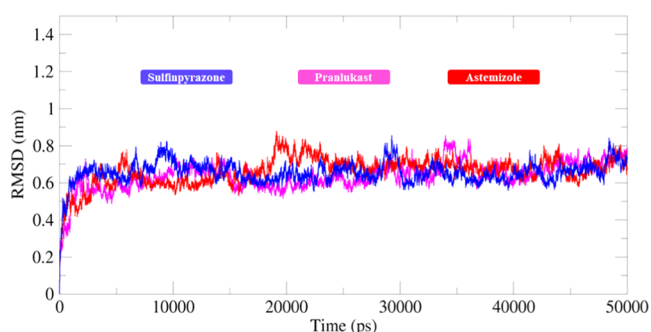
no.	drugs	functions	repositioned functions
1	chlorthalidone	hypertension	osteosarcoma of bone in childhood
2	astemizole	allergy	Ewing sarcoma
3	ketoconazole	seborrheic dermatitis	osteosarcoma of bone in children
4	sulfapyrazone	gouty arthritis	Ewing sarcoma
5	pranlukast	allergic rhinitis and asthma	Ewing sarcoma

**3.8. Molecular Dynamics Simulation.** Based on molecular docking, pharmacogenomics, and literature mining results, sulfapyrazone, chlorthalidone, and astemizole docked structures were selected to evaluate the residual flexibility in the target protein. The MD simulation study was employed at 50 000 ps using Gromacs 4.5.4. tool to generate root mean square deviations (RMSDs), root mean square fluctuations (RMSFs), solvent-accessible surface area (SASA), and radius of gyration ( $R_g$ ) graphs.

**3.8.1. Root Mean Square Deviation and Fluctuations.** The RMSD and RMSF graphs were generated to understand the protein backbone behavior in the simulation running time. The RMSD results showed that in docking structures, protein backbone deviation remained steady and stable with the passage of simulation time frame 0–50 000 ps. All of the graph lines exhibited an increasing trend with RMSD values ranging from 0.1 to 0.3 nm from 0 to 30 000 ps.

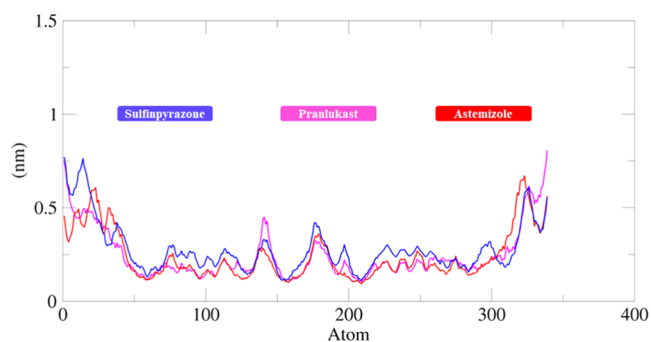
Initially, all of the graph lines (blue, pink, and red) of wild and both docked complexes showed an increasing trend with RMSD values of 0.1–0.7 nm from 0 to 5000 ps. After 5000–10 000 ps, the graph lines remained stable with a constant RMSD value at 0.6 nm. From 10 000 to 20 000 ps, all three graph lines remained stable; however, little fluctuation was observed in the astemizole structure (red), and the value increased to 0.8 nm. However, with the passage of simulation time, again stable behavior was attained from simulation time of 20 000–30 000 ps. After that, from 30 000 to 50 000 ps, all of the structures exhibited a stable constant RMSD value (0.6 nm). The overall RMSD graphs showed that both docked complexes and wild EWS protein remained stable and fluctuated less in the simulation time frame. The generated graph results showed stable behavior in the backbone of all protein complexes. The results showed that the binding of all of these drugs did not affect the structural configurations of EWS and remained stable during the simulation time (Figure 14).

The RMSF results of both docked complexes and standard EWS (blue, pink, and red curves) show that all residues dynamically fluctuated from N to C terminals. The protein structures are composed of 346–458 AA with different structural architectures. A couple of peaks have been observed



**Figure 14.** RMSD graph of sulfapyrazone–EWS (blue), chlorthalidone–EWS (pink), and astemizole–EWS (red) docked structures. In the generated graph, the Y-axis showed RMSD values, whereas the X-axis represents simulation time from 0 to 50 000 ps.

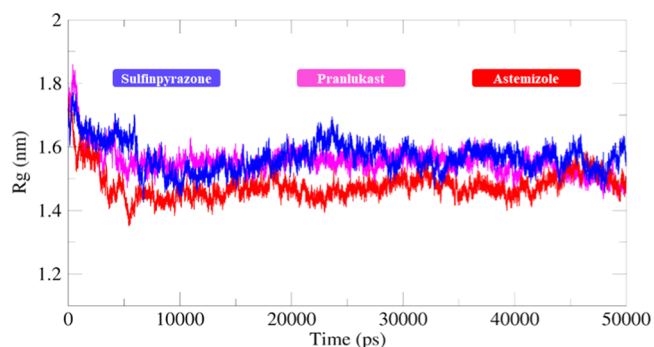
at both terminal regions. The remaining parts of all protein structures in Figure 14 remained stable throughout the simulation time (0–500 000 ps). Moreover, the central region of the protein, which consists of the binding pocket also showed few variations and fluctuations in the protein molecules. However, these variations do not cause much disturbance in the protein conformations, which ensures that our docking results are much more stable and steady in behavior. Residues from 346 to 401 AA showed fluctuations due to the loop region, whereas from 15 to 30 fewer fluctuations were observed with an RMSF value of 0.25 nm. Moreover, amino acids comprising  $\beta$ -sheets and  $\alpha$ -helices (401–445 AA) also remained stable in the simulation graph. After that, a couple of fluctuations peaks were observed from 445 to 460 AA. It has been observed that some interacting residues are also present in this region; after binding with the drug, they may disturb the protein structure, and the RMSF value increased from 0.25 to 0.5 nm. Moreover, from 460 to 490 AA, again smooth and steady peaks were seen, whereas after that again higher fluctuation peaks were observed due to the loop region of the Ewing sarcoma protein (Figure 15).



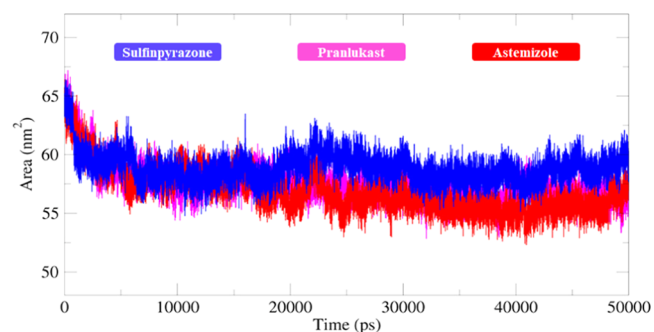
**Figure 15.** RMSF graph of sulfapyrazone–EWS, chlorthalidone–EWS, and astemizole–EWS docked structures. The Y-axis shows RMSF (nm) values, whereas X-axis represents residues of EWS.

**3.8.2. Solvent-Accessible Surface Area and Radius of Gyration.** The structural compactness of protein was calculated by the radius of gyration ( $R_g$ ). The generated results depicted that  $R_g$  values of all of the docked structures showed few variations from 1.5 to 1.7 nm. Initially, the graph lines were not much stable and showed few fluctuations from 0 to 5000 ps, while after that, stable behavior with few fluctuations was observed from 5000 to 10 000 ps time scale.

After that, no bigger fluctuations were observed in graph lines and the  $R_g$  value also remained stable at 1.6 nm. The solvent-accessible surface areas (SASAs) were also observed and are shown in Figure 16. The results showed that the values of SASA of all five docked complexes were centered on 60 nm<sup>2</sup> in the simulation time 0–50 000 ps (Figure 17).



**Figure 16.**  $R_g$  graph of sulfapyrazone–EWS, chlorthalidone–EWS, and astemizole–EWS docked complexes for a simulation time of 0–50 000 ps.



**Figure 17.** SASA graph of sulfapyrazone, chlorthalidone, and astemizole–EWS docked structures for a simulation time frame of 0–50 000 ps.

## 4. CONCLUSIONS

Drug repositioning is a computational approach employed for drug discovery. The current study evaluates the repositioning of known drugs for ES using shape-based screening, molecular docking pharmacogenomics, and MD simulation approaches. The computational shaped-based screening results showed that 100 FDA-approved drugs exhibited good structural similarity and scores with standard (pazopanib). Moreover, docking profile and pharmacogenomics evaluations depicted that from the bunch of 24 only five drugs were most active and showed good results compared to other drugs. The detailed pharmacogenomics and extensive data mining showed that three drugs have a direct association with ES by targeting different genes. Moreover, MD simulation results also exposed that these three drugs showed better profiles with respect to their RMSD, RMSF, SASA, and  $R_g$  evaluations graphs and steadily stable behavior was observed in all docking complexes. Taken together, it has been concluded that predicted astemizole, sulfapyrazone, and pranlukast exhibited better repositioning profiles as compared to other screened FDA-approved drugs. Therefore, sulfapyrazone, pranlukast, and astemizole may be potentially used in the treatment of ES after *in vitro* and clinical assessment in the future.



## ■ ASSOCIATED CONTENT

### SI Supporting Information

The Supporting Information is available free of charge at <https://pubs.acs.org/doi/10.1021/acsomega.2c00518>.

The authors have declared all of the raw data in the supplementary file; furthermore, no paid software and tools were employed and only free available tools were used to evaluate the results (PDF)

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### Author Contributions

M.H. designed this study and supervised in experiments; M.H. wrote the initial draft of the manuscript; M.Y. and S.S. collected data and performed experiments; and A.K. edited the manuscript and compiled it into the final format.

### Notes

The authors declare no competing financial interest.

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