



Supporting Information

# Combined Pharmacophore and Grid Independent Molecular Descriptors (GRIND) Analysis to Probe 3D features of Inositol 1, 4, 5-trisphosphate Receptor (IP<sub>3</sub>R) Inhibitors in Cancer

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## Supporting Information

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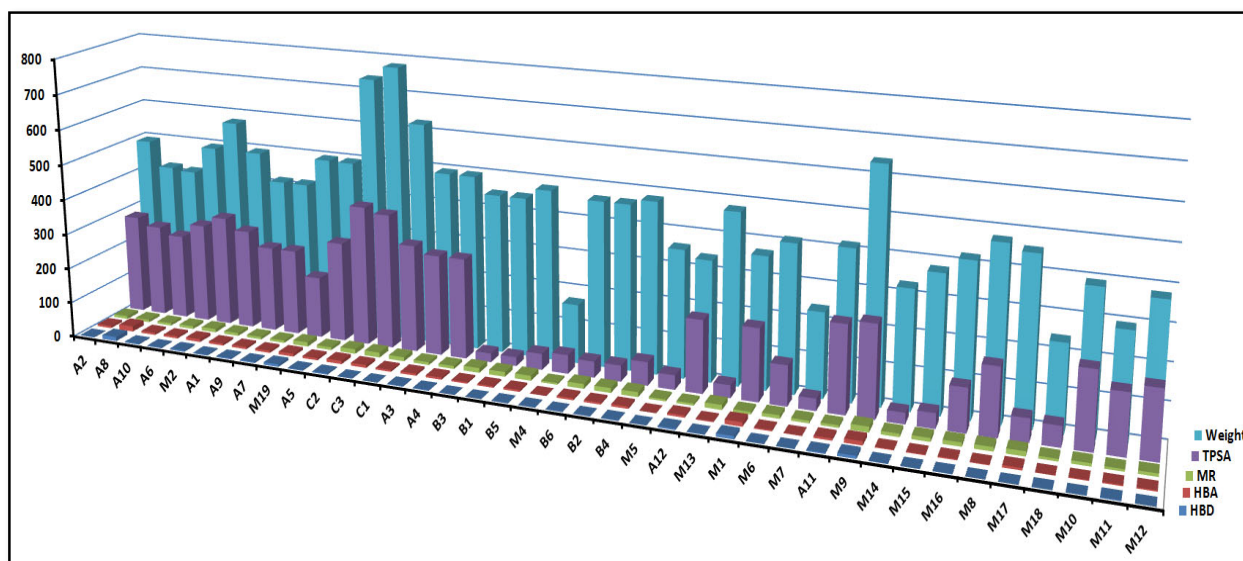
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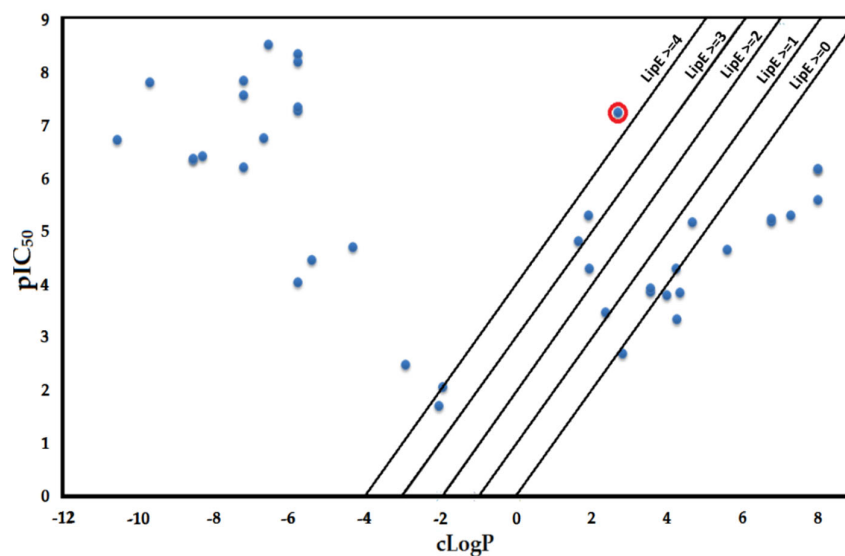
## 1. Results

**Table S1.** The activity landscape and LipE profile of the IP3R ligand dataset.

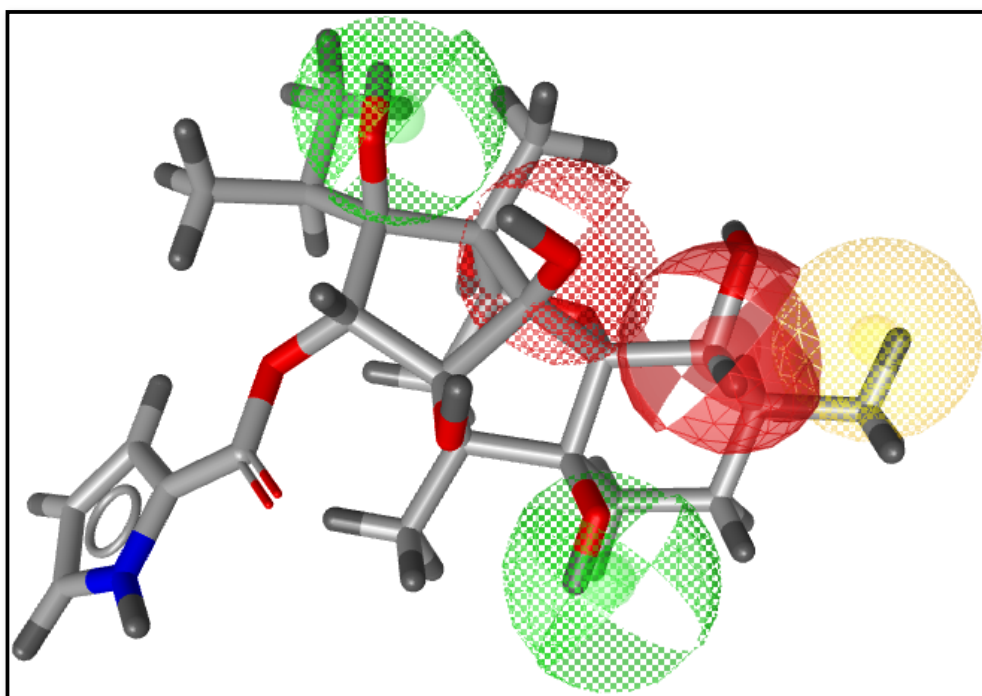
Comp	IC <sub>50</sub> (μM)	logP(o/w)	cLogp	pIC50	LipE	Ref
A2	0.02	-7.5	-7.2	1.8	15.1	[1]
A8	0.04	-6.2	-5.8	0.4	13.1	[1]
A10	0.01	-6.6	-5.7	1.9	13.9	[2]
A6	0.43	-7.7	-8.5	0.2	14.9	[2]
M2	0.02	-4.8	-7.2	7.5	17.5	[3]
A1	0.03	-7.5	-7.2	1.6	14.8	[2]
A9	0.62	-7.7	-7.2	1.3	13.4	[2]
A7	3.01	-6.4	-5.8	2.2	14.1	[1]
<b>M19</b>	<b>0.05</b>	<b>1.5</b>	<b>2.71</b>	<b>6.7</b>	<b>4.5</b>	<b>[4]</b>
A5	0.17	-7.5	-6.7	0.7	13.4	[1]
C2	0.19	-2.8	-6.1	6.7	17.2	[5]
C3	0.38	-3.9	-8.2	6.4	14.7	[5]
C1	0.42	-1.2	-4.2	6.3	14.9	[5]
A3	0.05	0.05	0.05	0.05	0.05	[2]
A4	-6.4	-6.4	-6.4	-6.4	-6.4	[1]
B3	5.86	6.5	6.8	5.2	-1.5	[6]
B1	6.60	5.7	4.7	5.2	0.5	[7]
B5	2.53	7.3	8.1	5.6	-2.4	[6]
M4	5.00	-0.6	1.9	5.3	3.3	[8]
M3	5.00	1.5	7.2	5.3	-1.9	[9]
B6	0.65	7.3	8.0	6.2	-1.8	[6]
B2	5.01	6.8	7.2	5.3	-1.9	[10]
B4	6.40	6.3	6.8	5.2	-1.5	[7]
M5	15.0	2.3	1.6	4.8	3.2	[11]
A12	20.0	-5.5	-4.3	-0.5	9.1	[12]
M13	22.0	6.1	5.6	4.6	-0.9	[13]
M1	34.0	-2.9	-5.4	4.4	9.9	[14]
M7	50.0	2.2	1.9	4.3	2.4	[8]
M6	50.0	1.9	4.2	4.3	0.1	[15]
A11	93.0	-6.9	-5.8	-1.3	9.8	[12]
M9	120	0.9	3.5	3.9	0.3	[16]
M14	140	3.1	3.5	3.8	0.3	[13]
M15	145	4.6	4.3	3.8	-0.5	[13]
M16	160	3.1	4.0	3.7	-0.2	[13]
M8	340	1.1	2.3	3.4	1.1	[16]
M17	450	5.3	4.2	3.3	0.9	[13]
M18	2000	1.7	2.8	2.6	-0.1	[4]
M10	3300	-2.5	-2.9	2.4	5.4	[16]
M11	8700	-1.9	-1.9	2.0	3.9	[16]
M12	20000	-2.3	-2.0	1.6	3.7	[16]



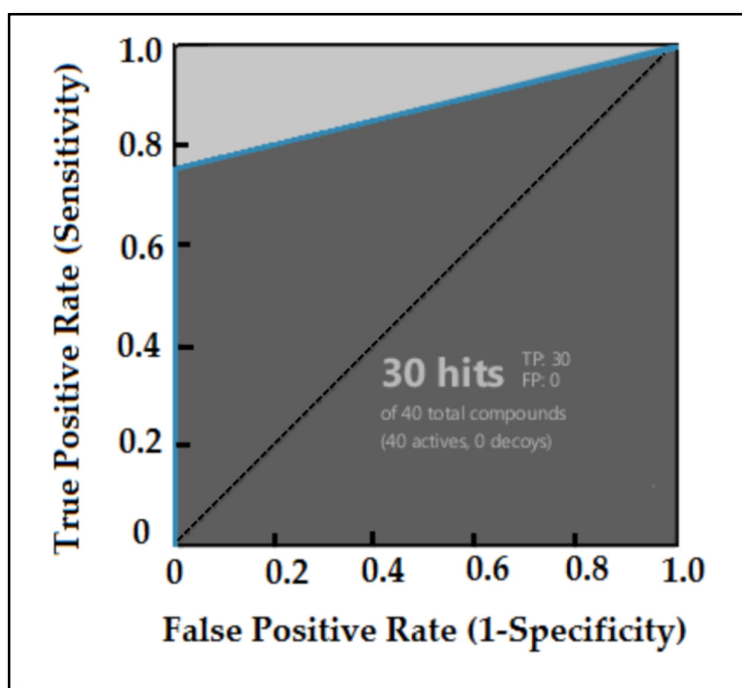
**Figure S1.** The physicochemical (molecular weight, TPSA, molar refractivity, and hydrogen bond acceptor/donor) properties of the IP<sub>3</sub>R ligand dataset.



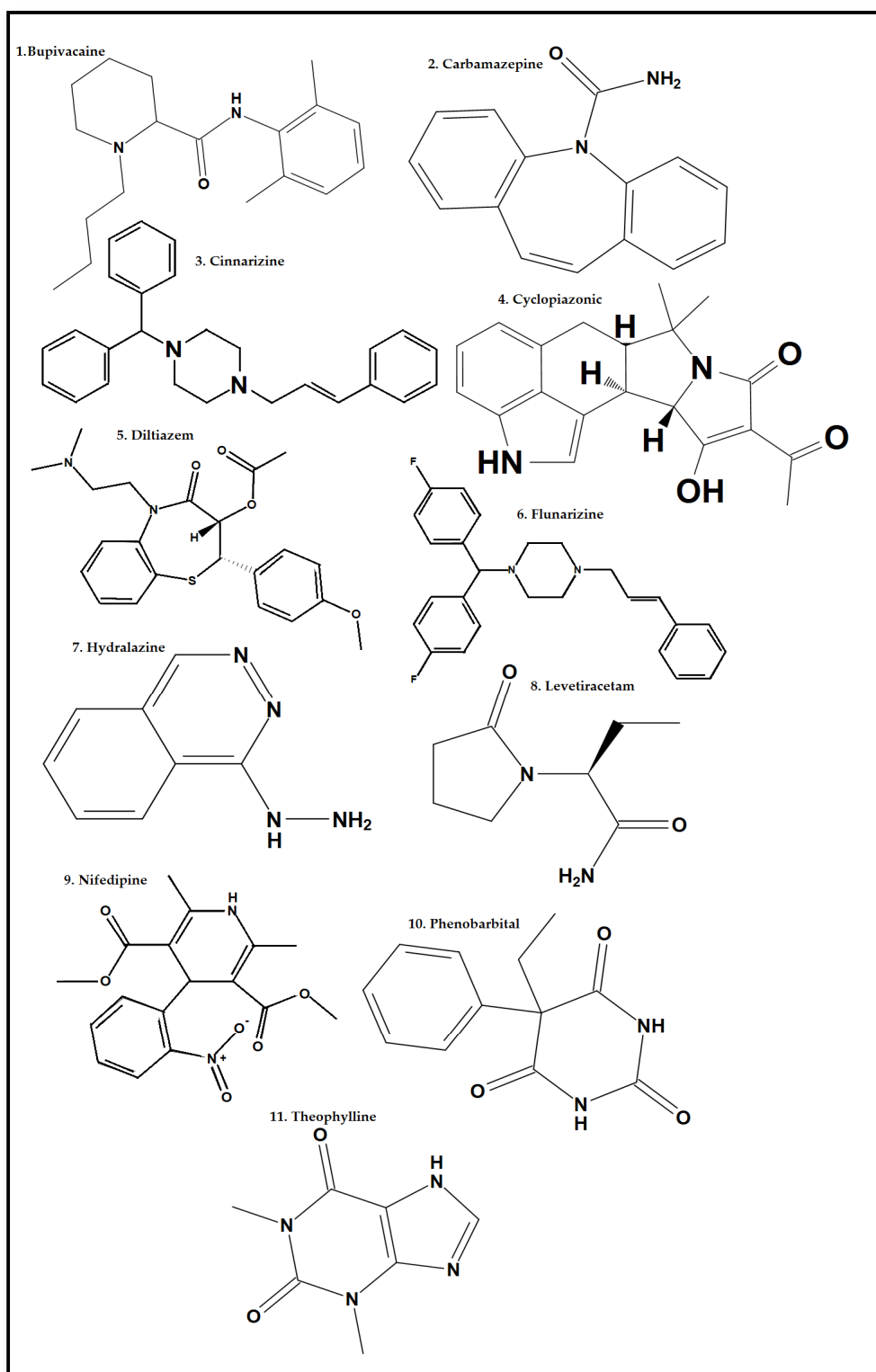
**Figure S2.** A plot of pIC<sub>50</sub> (inhibitory Potency) versus clogP showing Lipophilic Efficiency (LipE) profile of IP<sub>3</sub>R inhibitors. M<sub>19</sub> (Ryanodine) is circled red and selected as a template molecule because of the lipophilic efficiency profile (the most potent compound in the dataset (IC<sub>50</sub>: 0.055  $\mu$ M) with a clogP value of 2.71 and LipE value of 4.6).



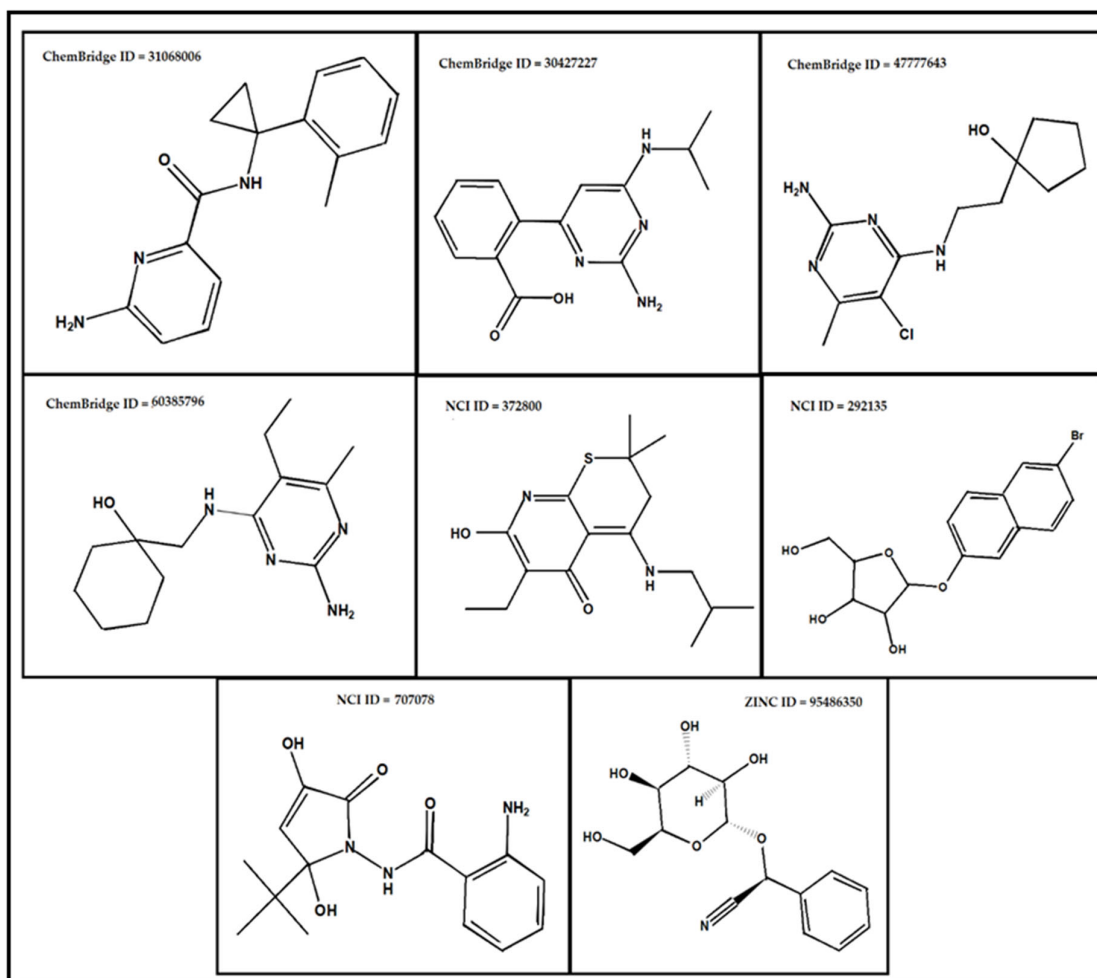
**Figure S3:** Shows the chemical features of the pharmacophore model responsible for the activity of ryanodine. The yellow circle represents the hydrophobic region. The hydrogen bond acceptor and hydrogen bond donors are represented by red and green circles respectively.



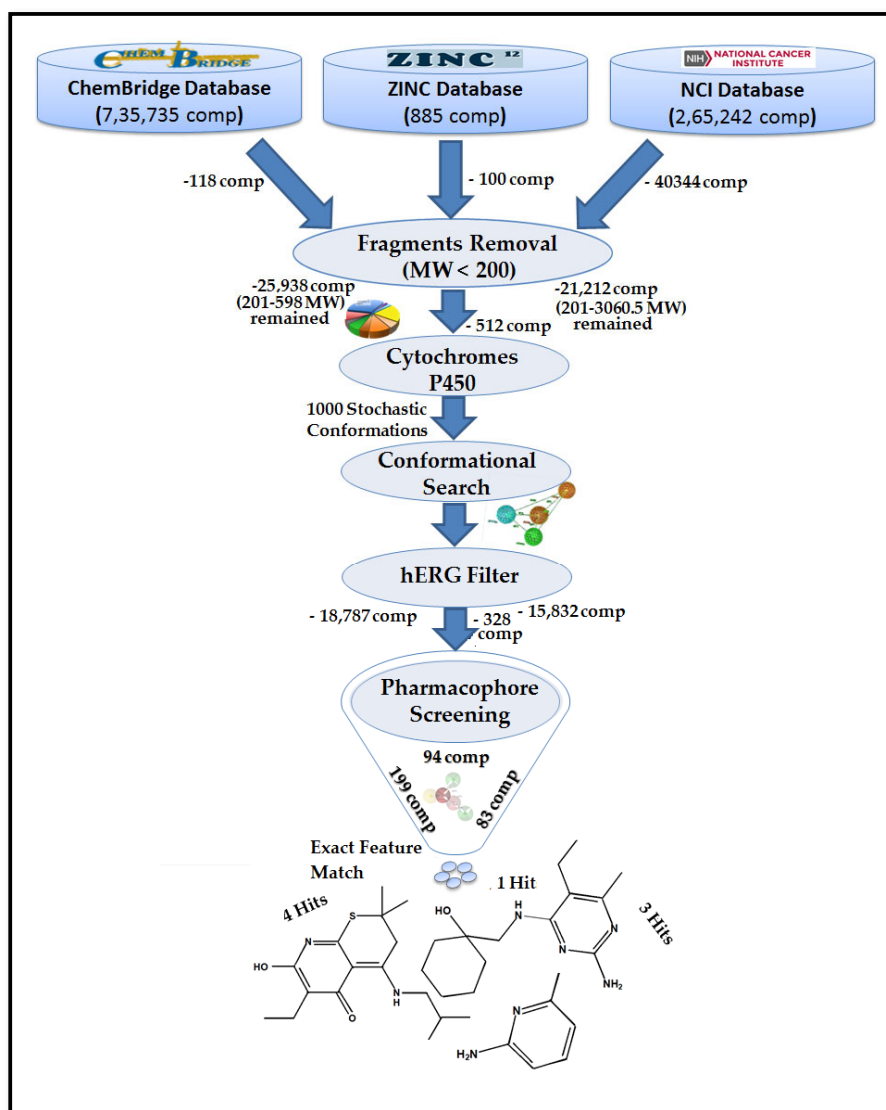
**Figure S4.** ROC curve between true positive (TP) rates (sensitivity) vs. false positive rate (1-specificity) of final selected pharmacophore model. Overall, out of 33 active compounds 30 were predicted as actives (TP) and 3 were predicted as inactive (FN).



**Figure S5.** The chemical structures of the compounds of the external test set used to validate the pharmacophore model. This test set is named as 'Blind set', as the  $IC_{50}$  values were not defined in the literature [17-19].

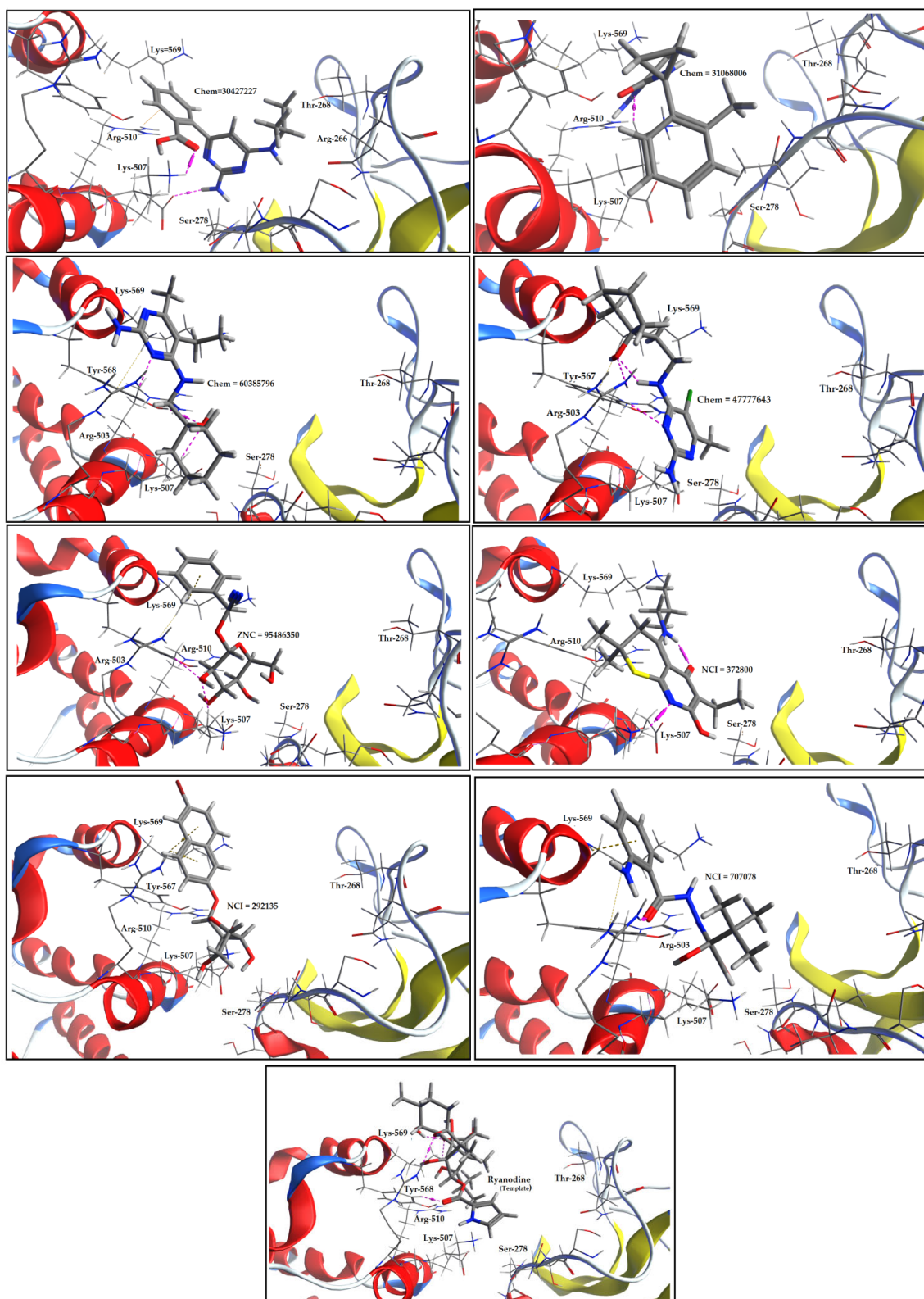


**Figure S6.** The chemical (2D) structures of the potential hit compounds shortlisted after pharmacophore based virtual screening of National Cancer Institute (NCI) database, ZINC database, and ChemBridge database.



**Figure S7.** A step by step protocol of the ligand based virtual screening. The 735735 compounds from ChemBridge database, 885 (natural) compounds from Zinc database, and 265242 compounds from the National Cancer Institute (NCI) database were screened. After several filters application and Pharmacophore model screening the 4 hits from ChemBridge, 4 hits from NCI and 2 hits from Zinc database were shortlisted as IP<sub>3</sub>R modulators (antagonists).



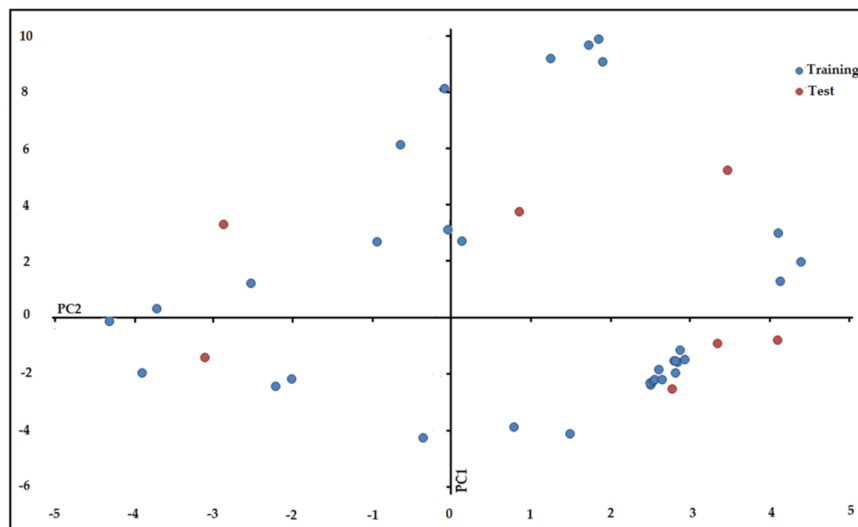


**Figure S8.** The best docked poses of shortlisted hits within the binding pocket of protein. The IP<sub>3</sub>R<sub>3</sub> protein is shown in backbone secondary structure and the interacting residues are shown in stick representation. Mostly, the ligand molecules interacted with Lys-569 and Lys-507 forming a  $\pi$ - $\pi$  interaction or surface contacts. Ligands interacted with Arg-503 and Arg-510 via hydrogen bond acceptor and donor interactions.

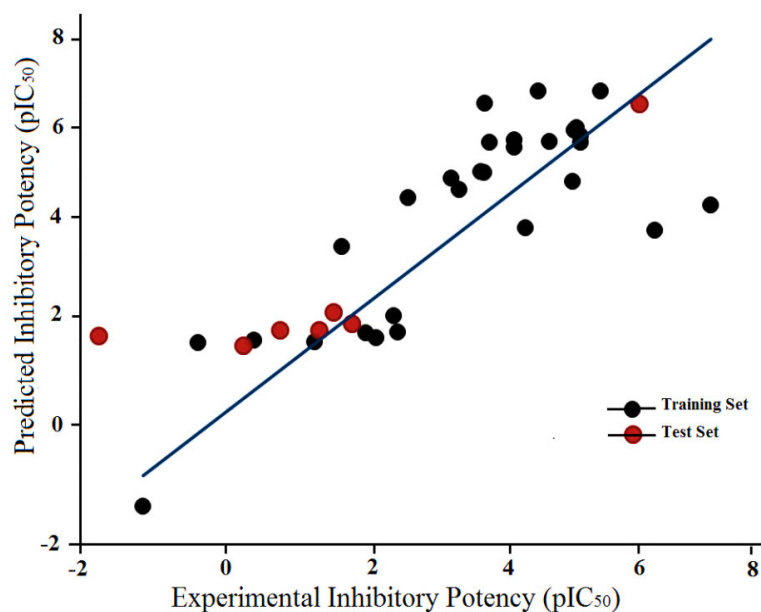


### 1. Principal Component Analysis:

The principal component analysis (PCA) [20] was performed to determine the structural variance in the training dataset by computing the complete GRIND descriptors set. In the training data set, 40 % structural variance has been described by the first two principal components (PC1 & PC2) (figure S9). The compounds in the form of a cluster at the right bottom side showed the molecules with a small structure containing only one hydrophobic ring, mainly class 'A' compounds and some from the class 'M'. Overall, the other compounds are more diverse with elongated chemical scaffolds and larger molecular weight, describing the data diversity.



**Figure S9.** A PCA plot between the first two principal components (PC1 & PC2) defining the descriptor space of the training set and test set.



**Figure S10.** Representing a correlation plot between experimental versus predicted inhibitory potencies (pIC<sub>50</sub>) of IP<sub>3</sub>R antagonists. The training set is represented by black circles while the test set is represented by red circles.

**Table S2.** The statistical parameters of PLS models generated by GRIND measured by cross-validation method.

Cross-Validation Method	Fractional Factorial Design (FFD) Cycle								
	Complete			FFD <sub>1</sub>			FFD <sub>2</sub>		
	Q <sup>2</sup>	R <sup>2</sup>	SDEP	Q <sup>2</sup>	R <sup>2</sup>	SDEP	Q <sup>2</sup>	R <sup>2</sup>	SDEP
LOO	0.61	0.64	1.1	0.68	0.71	1.0	<b>0.70</b>	<b>0.72</b>	<b>0.9</b>
LFO (1-5)	0.60	0.62	1.2	0.62	0.63	1.1	<b>0.64</b>	<b>0.66</b>	<b>1.0</b>
LFO (5-10)	0.62	0.64	1.1	0.64	0.65	1.0	<b>0.62</b>	<b>0.65</b>	<b>1.1</b>
LFO (10-15)	0.59	0.61	1.3	0.59	0.60	1.2	<b>0.59</b>	<b>0.61</b>	<b>1.3</b>
LFO (16-20)	0.58	0.60	1.2	0.59	0.61	1.1	<b>0.60</b>	<b>0.61</b>	<b>1.1</b>
LFO (21-25)	0.59	0.62	1.4	0.58	0.61	1.3	<b>0.60</b>	<b>0.62</b>	<b>1.2</b>
LFO (26-30)	0.61	0.63	1.2	0.61	0.60	1.3	<b>0.60</b>	<b>0.61</b>	<b>1.3</b>

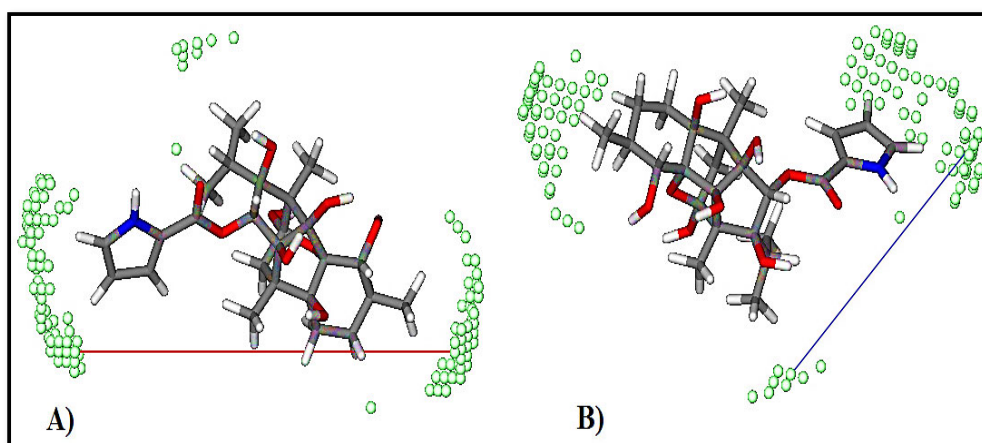
LOO = Leave-One-Out, LFO = Leave-Five-Out

**Table S3.** The experimental inhibitory potency compared with inhibitory potency predicted by GRIND. The residual values of  $\pm 2$  log units are considered optimal.

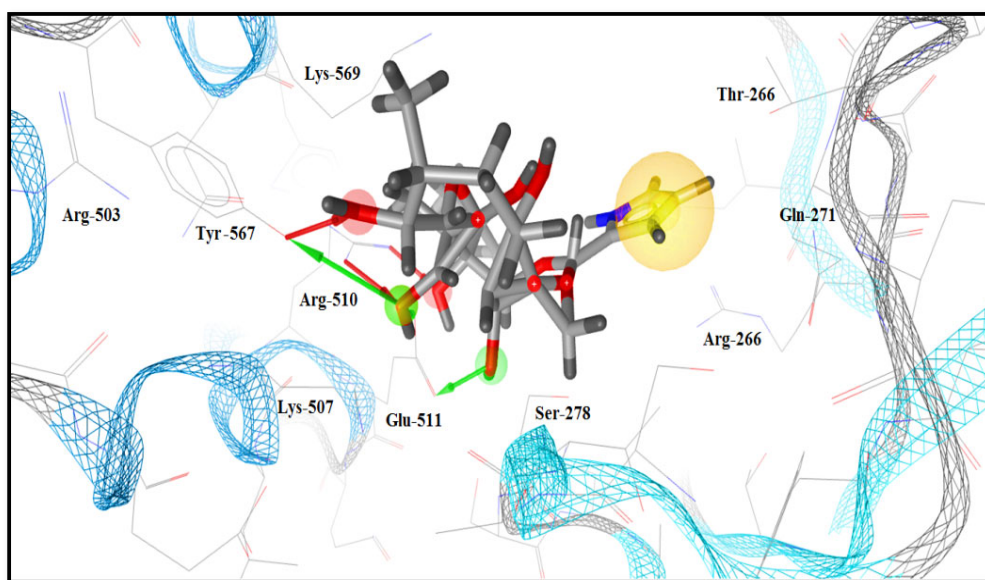
Comp	Experimental pIC <sub>50</sub>	Predicted pIC <sub>50</sub>	Residual Value	rm <sup>2</sup>	S <sub>new</sub>	AD (Outlier)	Comp	Experimental pIC <sub>50</sub>	Predicted pIC <sub>50</sub>	Residual Value	rm <sup>2</sup>	S <sub>new</sub>	AD (Outlier)
Training Set							Training Set						
A <sub>4</sub>	2.53	1.24	1.28	-0.33	1.29	-	M <sub>6</sub>	4.3	4.36	-0.06	3.24	2.59	-
A <sub>8</sub>	0.36	1.12	-0.76	0.46	0.96	-	M <sub>7</sub>	4.3	4.24	0.05	3.33	0.8	-
A <sub>3</sub>	1.28	1.09	0.18	0.73	2.61	-	A <sub>1</sub>	1.57	1.56	0.01	1.43	1.12	-
A <sub>12</sub>	-0.47	1.08	-1.55	0.11	0.93	-	M <sub>9</sub>	3.92	4.32	-0.4	1.44	1.52	-
A <sub>7</sub>	2.21	1.16	1.05	-0.05	0.96	-	C <sub>2</sub>	6.7	4.9	1.8	-2.28	1.18	-
A <sub>11</sub>	-1.3	-1.56	0.26	-0.63	0.59	-	A <sub>5</sub>	0.76	1.28	-0.52	0.21	1.69	-
B <sub>5</sub>	5.59	5.15	0.43	1.92	1.21	-	A <sub>6</sub>	0.2	1.03	-0.82	0.01	1.31	-
B <sub>4</sub>	5.19	4.51	0.67	0.94	1.1	-	C <sub>1</sub>	6.3	4.46	1.84	-2.24	1.61	-
B <sub>1</sub>	5.18	3.69	1.48	-1.12	1.13	-	M <sub>13</sub>	4.65	5.15	-0.49	1.39	1.53	-
B <sub>2</sub>	5.3	4.43	0.86	0.38	1.15	-	M <sub>14</sub>	3.85	4.95	-1.1	-0.18	1.47	-
B <sub>3</sub>	5.23	4.56	0.66	0.98	1.34	-	M <sub>15</sub>	3.83	3.84	-0.01	3.44	0.79	-
C <sub>3</sub>	6.42	2.9	3.51	-5.6	1.58	-	M <sub>16</sub>	3.79	3.85	-0.06	2.86	0.89	-
M <sub>1</sub>	4.46	2.93	1.53	-1.05	1.04	-	A <sub>2</sub>	1.85	1.38	0.46	0.59	1.19	-
M <sub>4</sub>	5.3	4.32	0.97	0.08	2.49	-	A <sub>9</sub>	1.35	1.28	0.07	0.99	1.07	-
M <sub>5</sub>	4.82	4.33	0.48	1.48	1.82	-	M <sub>19</sub>	7.25	5.53	1.72	-2.25	1.69	-

**Table S4.** The experimental inhibitory potency of test set compared with inhibitory potency predicted by GRIND. The modified  $r^2$  ( $rm^2$ ) calculated and the values greater than 0.5 are considered optimal.

Comp	Experi- mental pIC <sub>50</sub>	Predicted pIC <sub>50</sub>	Residual Value	$rm^2$	$S_{new}$	AD_Info (Outlier)
Test Set						
M <sub>8</sub>	3.46	3.55	-0.09	2.434	1.29	-
M <sub>11</sub>	2.06	1.24	0.81	0.195	1.13	-
M <sub>12</sub>	1.69	2.63	-0.93	0.576	1.16	-
M <sub>10</sub>	2.48	1.51	0.96	0.464	1.72	-
M <sub>17</sub>	3.34	3.74	-0.4	1.222	1.42	-
M <sub>18</sub>	2.69	3.42	-0.72	0.393	0.92	-
A <sub>10</sub>	1.96	1.18	0.78	0.596	0.54	-
B <sub>6</sub>	6.18	4.94	1.23	0.966	2.29	-



**Figure S11.** (A). TIP contour around the template molecule showing the curved molecular boundary at a wider distance of 16.40 Å - 16.80 Å is positively correlated with the inhibitory potency of IP<sub>3</sub>R. (B). whereas, the linear formed TIP at a shorter distance of 10.00 Å - 10.40 Å is negatively correlated with the inhibitory potency of IP<sub>3</sub>R.



**Figure S12.** The binding pose of template molecule representing important pharmacophoric features in complementing with amino acid residues within IP<sub>3</sub>R binding core.

## 2. Materials and Methods

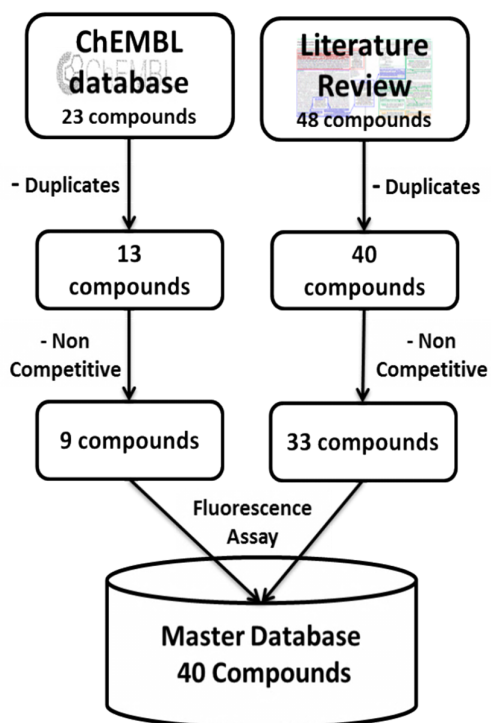


Figure S13. Step-by-step data curation process to obtain a master database of IP<sub>3</sub>R.

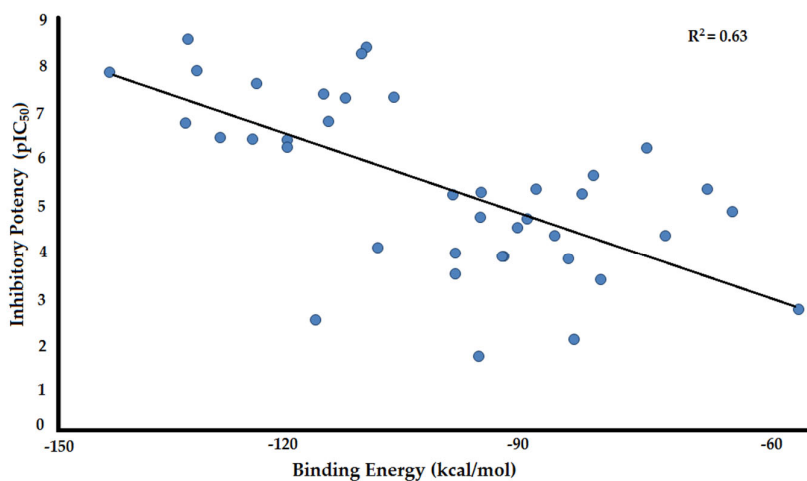


Figure S14. A correlation plot between binding energies of top docked poses vs. potential inhibitory potency (pIC<sub>50</sub>) showing good correlation i.e. 0.63.

## 2. Conformational Analysis of Ligand dataset for GRIND:

### 1. Energy minimized conformations

Briefly, a stochastic search algorithm in MOE 2019.01 [21] was applied to generate energy minimized conformations of the ligand dataset. The generated conformations were ranked according to their energy values and a total of 300 conformations were produced. Each ligand with the lowest energy score was considered for the GRIND analysis.

### 2. Standard 3D conformations

To obtain standard 3D conformations of the ligand dataset, an online version of CORINA software [22] was used. The 3D model of a molecule is built-in CORINA by connecting the mono-centric fragments with standard bond lengths and bond angles. The dihedral angles along with torsion angles of ring systems and the Van der Waals and electrostatic (non-bonded) interactions are also considered and minimized. The final 3D conformation of each ligand was further subjected to Pentacle v 1.0.7 [23] as input for GRIND analysis.

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