



VinaLigGen: A method to generate Ligplots and retrieval of hydrogen and hydrophobic interactions from protein-ligand complexes

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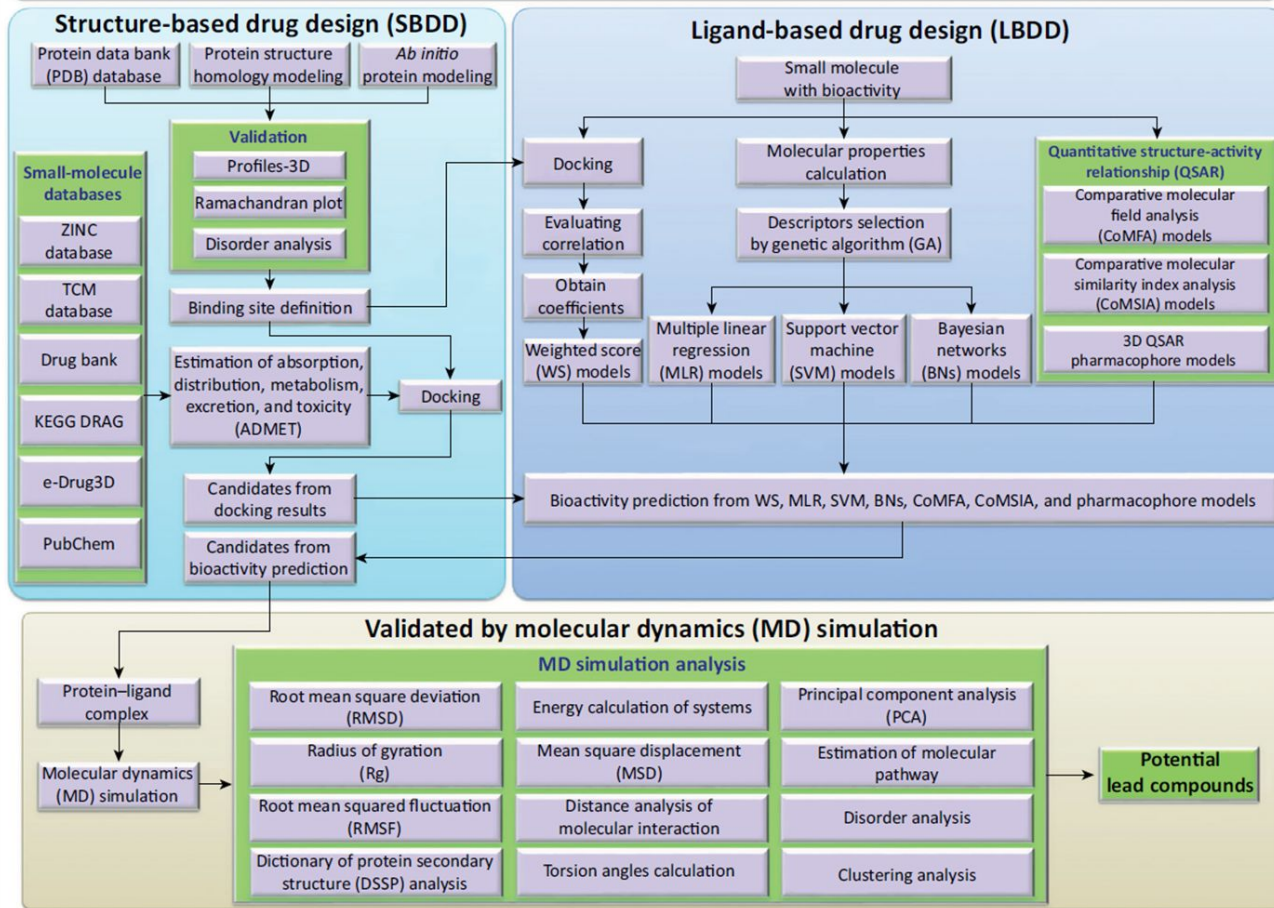
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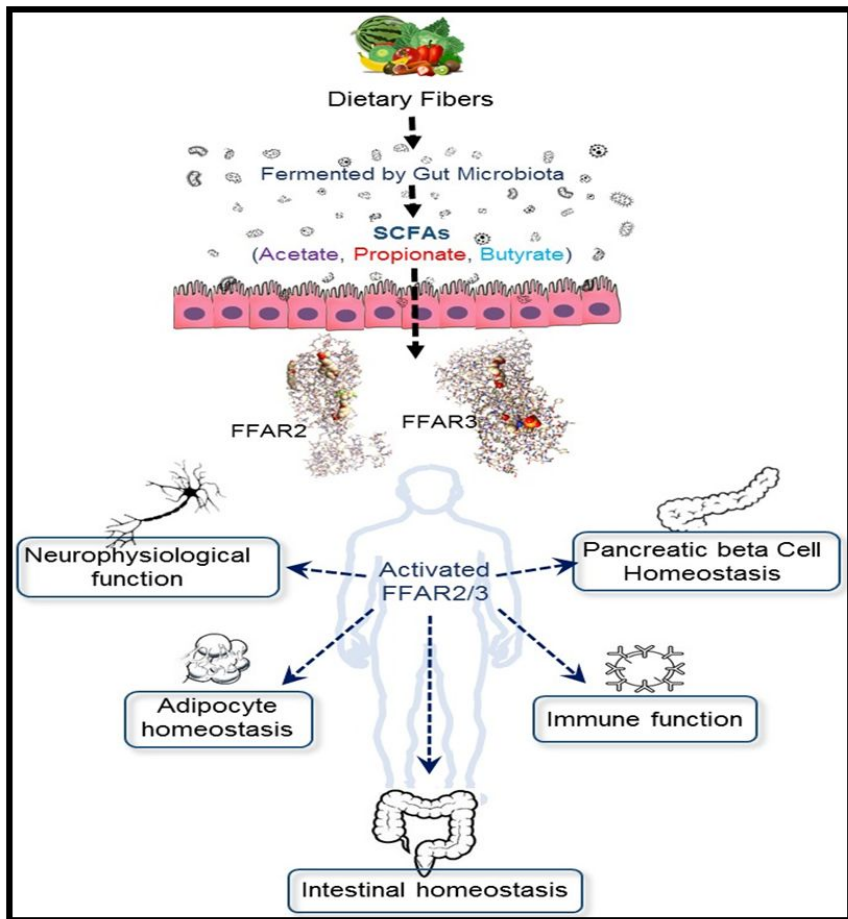
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18 September 2023

Flowchart of computer-aided drug design



Introduction to the problem



| S.No. | Name of Libraries | No. of Compounds |
|--------------|---------------------------------------|------------------|
| 1 | AfroDb Natural Products | 1008 |
| 2 | AnalytiCon Discovery NP | 20000 |
| 3 | Herbal Ingredients In-Vivo Metabolism | 1465 |
| 4 | Herbal Ingredients Targets | 9862 |
| 5 | IBScreen NP | 68000 |
| 6 | Indofine Natural Products | 20000 |
| 7 | Nubbe Natural Products | 2147 |
| 8 | Specs Natural Products | 800 |
| 9 | TCM Database @ Taiwan | 20000 |
| 10 | NPACT Database | 1574 |
| Total | | 144356 |

Vina score and LigPlot generation

Manual procedure

Number of ligands 16

Number of conformations/ligand is

= $16 * 9$ (docked vina conformations)

= 144 conformations

To generate LigPlot of 144 conformations

= $144 * 2$ minutes/conformation

= 288 minutes

= ~5 Hours

Our solution:

20 - 30 minutes

| ZINC ID | Compound Name | Human FFAR2 |
|-----------------|----------------------------------|-------------|
| ZINC00968029 | Darwinol | -6.6 |
| ZINC02040990 | Beta-terpineol | -6.5 |
| ZINC95099135 | 7-hydroxycamphene | -6.1 |
| ZINC01597139 | (±)-Carvomenthol | -6.1 |
| ZINC00157548 | Norpseudoephedrine hydrochloride | -6 |
| ZINC00403588 | Synephrine | -6 |
| ZINC00388198 | Octopamine hydrochloride | -5.8 |
| ZINC02034811 | 3-Pinanone | -5.8 |
| ZINC00968099 | Borneol | -5.8 |
| ZINC00074836 | Ephedrine | -5.7 |
| ZINC00967571 | (+)-Fenchone | -5.7 |
| ZINC01081099 | FENCHOL | -5.5 |
| ZINC03581377 | L-Leucinamide hydrochloride | -5.4 |
| ZINC00895132 | Butyrate | -4.3 |
| PubChem CID:175 | Acetate | -3 |
| ZINC118616157 | CATPB | -9.8 |

VinaLigGen - AutoDock Vina results to LigPlot Generation



- A program that extracts hydrogen, hydrophobic interactions and generates postscripts/png of these interactions for better analysis.
- The tool takes the docked protein-ligand complex (_out.pdbqt files) along with the macromolecule (.pdb) file.
- Before extracting interaction data, It shortlists docked complexes based on the given molecule IDs.
- This tool does not require any pre-processing of the complexes.
- The output of this tool is a .zip file which consists of,
 - An excel sheet which has the hydrogen and hydrophobic bond data.
 - A folder 'Target Files', Which has the files that were considered for generating the above.
 - Another folder 'LigPlot', Which has the postscript, and bond interactions data files.

VinaLigGen - AutoDock Vina results to LigPlot Generation

- The docked complexes used as input files can be of the following 3 types,
 - Type 1: The Molecule ID is in the file name itself.
 - Type 2: The Molecule ID is inside the file.
 - Type 3: This one is very similar to Type 2, But the only difference is that the molecule file is inside a folder and the molecule ID is inside the file. As per this example, the molecule i.e 'out.pdbqt' is placed inside. Example: 'KUS-LIB1_ligand_001' and the molecule ID is inside out.pdbqt

```
KUS-LIB1_ligand_001.pdbqt PDBQT File  
  
REMARK Name = ZINC000000007677  
REMARK 0 active torsions:  
REMARK status: ('A' for Active; 'I' for Inactive)
```

Type 2

```
zinc000000007677_uff_e=1783.05_out.pdbqt PDBQT File
```

Type 1

```
KUS-LIB1_ligand_001 File folder  
  
out.pdbqt PDBQT File  
  
REMARK Name = ZINC000000007677  
REMARK 0 active torsions:  
REMARK status: ('A' for Active; 'I' for Inactive)
```

Type 3

VinaLigGen - AutoDock Vina results to LigPlot Generation



Upload Required Files

Upload Ligands



Drag and drop file here

Limit 200MB per file • ZIP

Browse files



ligands.zip 30.1KB



Upload Macro-Molecule



Drag and drop file here

Limit 200MB per file • PDB

Browse files



FFAR2_Human.pdb 217.8KB



Upload Target Molecule list



Drag and drop file here

Limit 200MB per file • TXT

Browse files



shortlisted.txt 33.0B



Upload

Select Type of Ligand files

- Type_1
 Type_2
 Type_3

Start!

Successfully uploaded

```
Zip files : ligands.zip
PDB File : macromolecule.pdb
Text File : shortlist.txt
```

Getting short listed files - using File name...

- Fetching short listed files...!
- SUCCESS: Fetched all the 3 short listed files!!

Starting Splitting of Files

- Fetched 3 files which are going to be split

Splitting Files...: 67% | 2/3 [00:07-00:03, 3.91s/it]

- SUCCESS: All the 3 files have been split to 27 files!!

Generating of ligplot Generator File!

- The files required are,
- ligplot_generator.bat
- hbadd.exe
- hbplus.exe
- ligplot.exe
- ligplot.prm

- Checking if all dependencies are satisfied...

- All dependencies satisfied!

- Please run the D:\Technical\VinaLigGen-main\Master_folder\top_40_files /ligplot_generator.bat file

Program has started

```
Processing ligand 175_1 |# [1/27]
Processing ligand 175_2 |### [2/27]
Processing ligand 175_3 |##### [3/27]
Processing ligand 175_4 |##### [4/27]
Processing ligand 175_5 |##### [5/27]
Processing ligand 175_6 |##### [6/27]
Processing ligand 175_7 |##### [7/27]
Processing ligand 175_8 |##### [8/27]
Processing ligand 175_9 |##### [9/27]
```

VinaLigGen - AutoDock Vina results to LigPlot Generation

Upload Ligands

Drag and drop file here
Limit 200MB per file • ZIP

ligands.zip 30.1KB

Upload Macro-Molecule

Drag and drop file here
Limit 200MB per file • PDB

FFAR2_Human.pdb 217.8KB

Upload Target Molecule list

Drag and drop file here
Limit 200MB per file • TXT

shortlisted.txt 33.0B

Upload

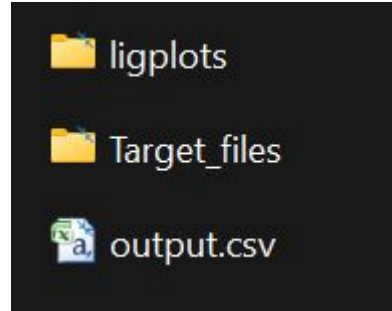
Select Type of Ligand files

Type_1
 Type_2
 Type_3

Start!

Download Files

Time Taken for the Job to Finish = 2 Minutes 1 Seconds



- hbadd.bonds
- hbadd.map
- hbdebug.dat
- hbplus.rc
- ligplot.bonds
- ligplot.drw
- ligplot.frm
- ligplot.hhb
- ligplot.nnb
- ligplot.pdb
- ligplot.png
- ligplot.ps
- ligplot.rcm
- ligplot.sum
- ranseed.dat
- ZINC01081099_5.hhb
- ZINC01081099_5.nnb
- ZINC01081099_5.pdb



VinaLigGen - AutoDock Vina results to LigPlot Generation

Significance of hydrogen bonding and hydrophobic interactions

Molecular Recognition: Hydrogen bonds enable specific interactions between ligands and receptors.

Binding Affinity: They contribute to strong non-covalent bonding and higher binding affinities.

Binding Orientation: Hydrogen bonds guide ligands into the correct binding position.

Binding Specificity: They help discriminate between similar ligands based on bonding patterns.

Scoring Functions: Docking algorithms use hydrogen bond data in scoring ligand-receptor complexes.

Drug Design: Hydrogen bonds inform the optimization of ligand structures for binding.

Biological Function: Hydrogen bonds are crucial in various biological processes.

Virtual Screening: Hydrogen bond predictions aid in screening potential drug candidates.

Complementarity: Hydrophobic interactions contribute to the complementarity between ligands and receptors.

Binding Affinity: They play a crucial role in stabilizing ligand-receptor complexes.

Nonpolar Regions: Hydrophobic groups on ligands and receptors interact favorably in nonpolar environments.

Core Binding: Hydrophobic interactions often form the core of protein-ligand complexes.

Specificity: They aid in discriminating between ligands with different hydrophobic properties.

Binding Strength: Hydrophobic interactions can significantly enhance the overall binding strength.

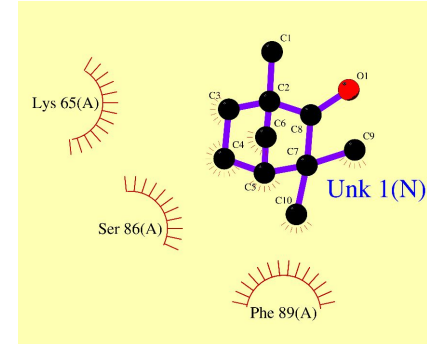
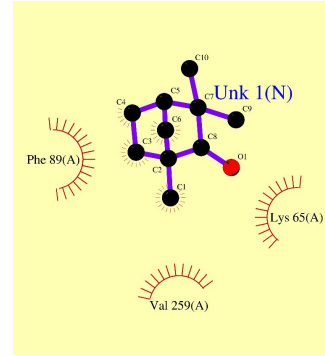
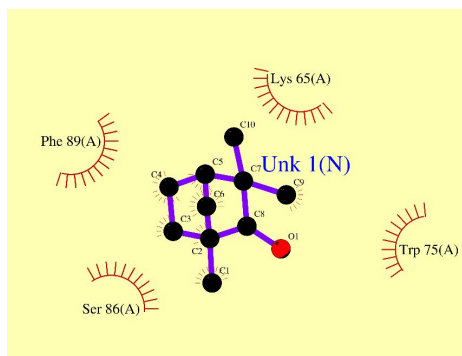
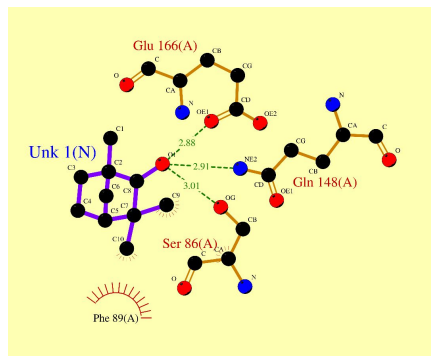
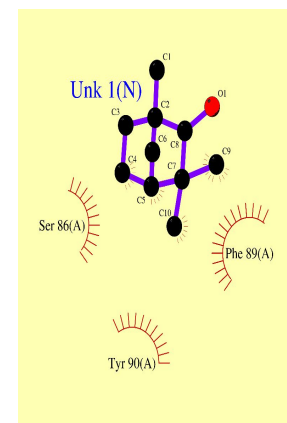
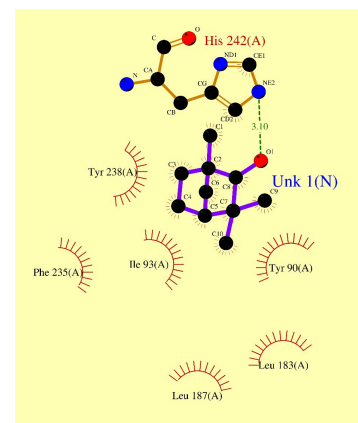
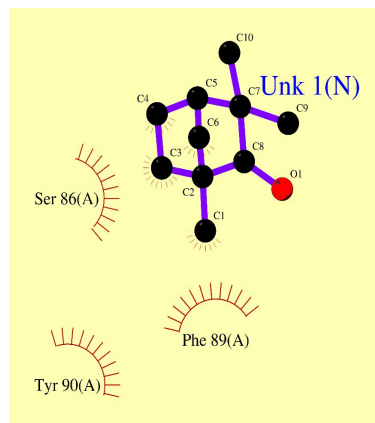
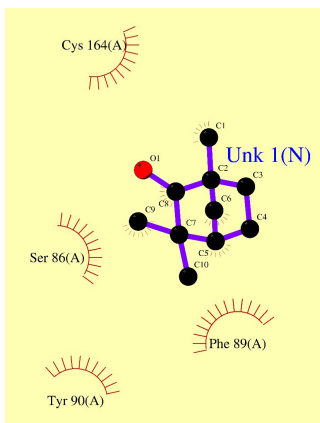
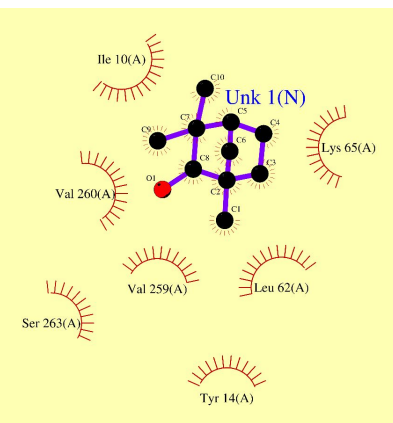
Structural Stability: They contribute to the structural stability of ligand-receptor complexes. 9



VinaLigGen - AutoDock Vina results to LigPlot Generation

| LIGAND_NAME | Hydrogen_bonds | d_distance | Hydrophobic_bonds | Hydrophobic_bond_distance |
|----------------|-------------------------|-------------|--|--|
| 175_1 | HIS242 | 2.88 | ARG180 PHE168 HIS242 VAL176 PHE168 | 3.78 3.83 3.87 3.78 3.78 |
| 175_2 | | | PHE231 SER262 PHE235 PHE231 ILE93 | 3.69 3.87 3.81 3.85 3.79 |
| 175_3 | | | LYS65 ILE10 | 3.77 3.74 |
| 175_4 | | | ILE17 TYR14 TYR14 | 3.85 3.62 |
| 175_5 | HIS242 | 3.24 | TYR90 TYR90 TYR90 TYR90 TYR90 | 3.83 3.87 3.85 3.79 |
| 175_6 | | | PHE235 | 3.46 3.56 3.60 3.78 3.80 3.78 |
| 175_7 | | | VAL260 LEU62 | 3.64 3.69 |
| 175_8 | | | SER86 THR85 THR85 THR85 LYS65 | 3.80 3.62 3.77 3.61 3.67 |
| 175_9 | | | THR85 THR85 THR85 | 3.89 3.76 3.73 |
| ZINC00388198_1 | TYR238 | 2.71 3.07 | TYR90 TYR90 TYR90 TYR90 LEU187 LEU183 | 3.78 3.66 3.84 3.81 3.48 3.65 3.52 |
| ZINC00388198_2 | ASN239 | 2.95 | TYR238 PHE235 HIS242 PHE235 PHE235 | 3.76 3.51 3.57 3.73 3.63 3.75 3.41 |
| ZINC00388198_3 | GLU166 | 3.03 2.98 | TRP75 TRP75 | 3.74 |
| ZINC00388198_4 | GLU166 | 3.04 3.00 | PHE89 SER86 | 3.68 |
| ZINC00388198_5 | TYR238 GLN148 | 2.93 2.90 | PHE89 PHE89 PHE89 PHE89 | 3.84 3.83 3.80 |
| ZINC00388198_6 | GLN148 SER86 GLU166 | 3.04 | SER86 THR85 PHE89 | 3.64 3.86 3.56 |
| ZINC00388198_7 | TYR238 THR85 | 2.70 3.06 | PHE89 PHE89 LYS65 | 3.66 3.62 |
| ZINC00388198_8 | GLU166 | 2.94 | SER86 SER86 | 3.70 |
| ZINC00388198_9 | TYR238 CYS164 | 2.88 3.33 | PHE89 PHE89 | 3.64 |
| ZINC01081099_1 | | | ILE10 ILE10 ILE10 VAL260 VAL259 VAL260 | 3.83 3.78 3.48 3.73 3.83 3.75 3.86 |
| ZINC01081099_2 | | | PHE89 PHE89 CYS164 | 3.84 3.78 |
| ZINC01081099_3 | | | TYR90 TYR90 PHE89 | 3.52 3.44 |
| ZINC01081099_4 | HIS242 | 3.1 | TYR90 TYR90 TYR90 TYR90 TYR90 LEU183 | 3.85 3.86 3.54 3.84 3.54 3.49 3.49 |
| ZINC01081099_5 | GLU166 GLN148 SER86 | 3.01 | PHE89 PHE89 PHE89 PHE89 SER86 | 3.83 3.45 3.72 3.65 3.56 |
| ZINC01081099_6 | | | PHE89 PHE89 PHE89 PHE89 LYS65 | 3.78 3.90 3.66 3.72 |
| ZINC01081099_7 | | | PHE89 PHE89 | 3.84 |
| ZINC01081099_8 | | | VAL259 PHE89 PHE89 PHE89 PHE89 VAL259 | 3.82 3.60 3.61 3.85 3.68 3.86 3.83 |
| ZINC01081099_9 | | | SER86 SER86 PHE89 PHE89 PHE89 PHE89 | 3.67 3.49 3.68 3.72 3.72 3.60 3.73 |

VinaLigGen - AutoDock Vina results to LigPlot Generation



Key

- Ligand bond
- Non-ligand bond
- Hydrogen bond and its length
- His 53 Non-ligand residues involved in hydrophobic contact(s)
- Corresponding atoms involved in hydrophobic contact(s)



VinaLigGen - AutoDock Vina results to LigPlot Generation

| Sl. No. | Compounds | H-Bonding Residues Human_FFAR2 | Probability as Agonist Ser86, Tyr90, His140, Ile145, Val179, Arg180, Leu183, Tyr238, His242, Arg255 |
|---------|---|--|--|
| 1 | 175 (Acetate) | His242, Tyr238 | Potential Agonist |
| 2 | ZINC0038819 8 (Octopamine hydrochloride) | Leu183, Tyr238, His242, Asn239, Gln148, Cys164, Glu166, Ser86, Glu68, Thr85, | 70% |
| 3 | ZINC0108109 9 (Fenchol) | Glu166, Ser86, His242, Gln148, Tyr238 | 50% |

ORIGINAL RESEARCH
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Activation of Microbiota Sensing – Free Fatty Acid Receptor 2 Signaling Ameliorates Amyloid- β Induced Neurotoxicity by Modulating Proteolysis-Senescence Axis

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Multiple emerging evidence indicates that the gut microbiota contributes to the pathology of Alzheimer’s disease (AD)—a debilitating public health problem in older adults. However, strategies to beneficially modulate gut microbiota and its sensing signaling pathways remain largely unknown. Here, we screened, validated, and established the agonists of free fatty acid receptor 2 (FFAR2) signaling, which senses beneficial signals from short chain fatty acids (SCFAs) produced by microbiota. The abundance of SCFAs, is often low in the gut of older adults with AD. We demonstrated that inhibition of FFAR2 signaling increases amyloid-beta (A β) stimulated neuronal toxicity. Thus, we screened FFAR2 agonists using an in-silico library of more than 144,000 natural compounds and selected 15 of them based on binding with FFAR2-agonist active sites. Fenchol (a natural compound commonly present in basil) was recognized as a potential FFAR2 stimulator in neuronal cells and demonstrated protective effects against A β -stimulated neurodegeneration in an FFAR2-dependent manner. In addition, Fenchol reduced AD-like phenotypes, such as A β -accumulation, and impaired chemotaxis behavior in *Caenorhabditis (C.) elegans* and mice models, by increasing A β -clearance via the promotion of proteolysis and reduced senescence in neuronal cells. These results suggest that the inhibition of FFAR2 signaling promotes A β -induced neurodegeneration, while the activation of FFAR2 by Fenchol ameliorates these abnormalities by promoting proteolytic A β -clearance and reducing cellular senescence. Thus, stimulation of FFAR2 signaling by Fenchol as a natural compound can be a therapeutic approach to ameliorate AD pathology.

Keywords: microbiota, free fatty acid, G-coupled protein receptor, FFAR2 (GPR43), fenchol, natural compounds, Alzheimer’s

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