

# Scaffold analysis in Python with RDKit and pandas

[Pandas](http://pandas.pydata.org/) (<http://pandas.pydata.org/>) is an open source, BSD-licensed library providing high-performance, easy-to-use data structures and data analysis tools for the Python programming language.

[RDKit](http://www.rdkit.org/) (<http://www.rdkit.org/>) is an open source chemistry toolkit.

```
In [20]: import pandas as pd
import rdkit.Chem as Chem
from rdkit.Chem import PandasTools
from rdkit.Chem import Draw
from rdkit.Chem import Descriptors
from rdkit.Chem.Draw import IPythonConsole # Enables RDKit IPython integration
```

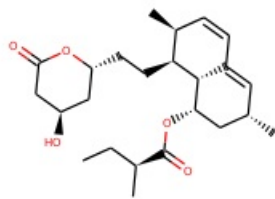
## IPython and RDKit

RDKit provides IPython integration

```
In [21]: mol = Chem.MolFromSmiles('O=C(O[C@@H]1[C@H]3C(=C/[C@H](C)C1)\C=C/[C@@H]([C@@H]3CC[C@H]2OC(=O)C[C@H](O)C2)C)[C@H](C)CC')
```

```
In [22]: mol
```

```
Out [22]:
```



```
In [23]: Descriptors.NumHDonors(mol)
```

```
Out [23]: 1
```

```
In [24]: Descriptors.MolLogP(mol)
```

```
Out [24]: 4.1955000000000004
```

## RDKit and pandas

Load 'approved drugs' downloaded from [www.drugbank.ca](http://www.drugbank.ca):

*% time is a ipython magic function that tells you how much time did certain operation take to finish. It will be used to give you a feeling about speed of certain functions*

```
In [25]: % time cpds = PandasTools.LoadSDF('approved.sdf', includeFingerprints=False)

CPU times: user 6.5 s, sys: 23.3 ms, total: 6.53 s
Wall time: 6.54 s
```

```
In [26]: cpds.columns
```

```
Out [26]: Index([u'ALOGPS_LOGP', u'ALOGPS_LOGS', u'ALOGPS_SOLUBILITY', u'BRANDS', u'CHEMICAL_FORMULA', u'DRUGBANK_ID', u'DRUG_GROUPS', u'EXACT_MASS', u'GENERIC_NAME', u'ID', u'INCHI_IDENTIFIER', u'INCHI_KEY', u'IUPAC_NAME', u'JCHEMA_ACCEPTOR_COUNT', u'JCHEMA_ACIDIC_PKA', u'JCHEMA_BASIC_PKA', u'JCHEMA_DONOR_COUNT', u'JCHEMA_LOGP', u'JCHEMA_PHYSIOLOGICAL_CHARGE', u'JCHEMA_POLARIZABILITY', u'JCHEMA_POLAR_SURFACE_AREA', u'JCHEMA_REFRACTIVITY', u'JCHEMA_ROTATABLE_BOND_COUNT', u'MOLECULAR_WEIGHT', u'SALTS', u'SMILES', u'SYNONYMS', u'ROMol'], dtype=object)
```

```
In [27]: len(cpds)
```

Out [27]: 1485

Assign the values of molnames and smiles (makes it easier to use this notebook on other sets with different col names)

```
In [28]: molnames = 'DRUGBANK_ID'  
        smiles = 'SMILES'
```

Keep only columns 'DRUGBANK\_ID', 'SMILES' and 'ROMol'

```
In [29]: cpds = cpds[[molnames, smiles, 'ROMol']]
```

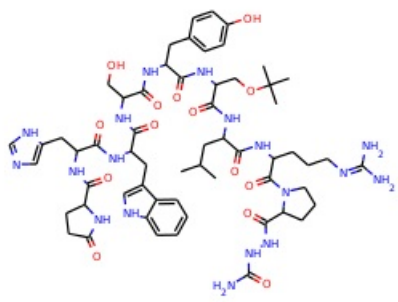
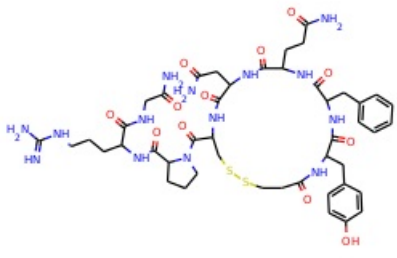
```
In [30]: cpds.columns
```

```
Out [30]: Index([u'DRUGBANK_ID', u'SMILES', u'ROMol'], dtype=object)
```

Look at two columns and only first 2 lines:

```
In [31]: cpds[[molnames, 'ROMol']].head(2)
```

Out [31]:

	DRUGBANK_ID	ROMol
0	DB00014	
1	DB00035	

Remove lines with NaN (empty) values and duplicates

```
In [32]: cpds = cpds.dropna()  
        cpds = cpds.drop_duplicates(molnames)  
        cpds = cpds.drop_duplicates(smiles)  
        len(cpds)
```

Out [32]: 1462

## Descriptors

Add some descriptors

```
In [33]: from rdkit.Chem import Descriptors
cpds['logp'] = cpds['ROMol'].map(Descriptors.MolLogP)
cpds['mw'] = cpds['ROMol'].map(Descriptors.MolWt)
```

Remove compounds with logp >= 5 and MW >= 500

```
In [34]: cpds = cpds[cpds['logp'] <= 5]
cpds = cpds[cpds['mw'] <= 500]
len(cpds)
```

Out[34]: 1143

```
In [35]: cpds[['molnames', 'logp', 'mw', 'smiles']].head()
```

Out[35]:

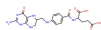
	DRUGBANK_ID	logp	mw	SMILES
6	DB00116	-0.2820	445.436	<chem>Nc1nc(=O)c2c([nH]1)NCC(CNc1ccc(C(=O)NC(CCC(=O)O)C(=O)O)cc1)N2</chem>
7	DB00117	-0.6359	155.157	<chem>NC(Cc1cnc[nH]1)C(=O)O</chem>
8	DB00118	-1.9222	399.453	<chem>C[S+](CCC(N)C(=O)O)CC1OC(n2cnc3c2ncnc3N)C(O)C1O</chem>
9	DB00119	-0.3400	88.062	<chem>CC(=O)C(=O)O</chem>
10	DB00120	0.6410	165.192	<chem>NC(Cc1ccccc1)C(=O)O</chem>

## Alternative visualisation of a table

Default takes a lot of space

```
In [36]: cpds.head(1)
```

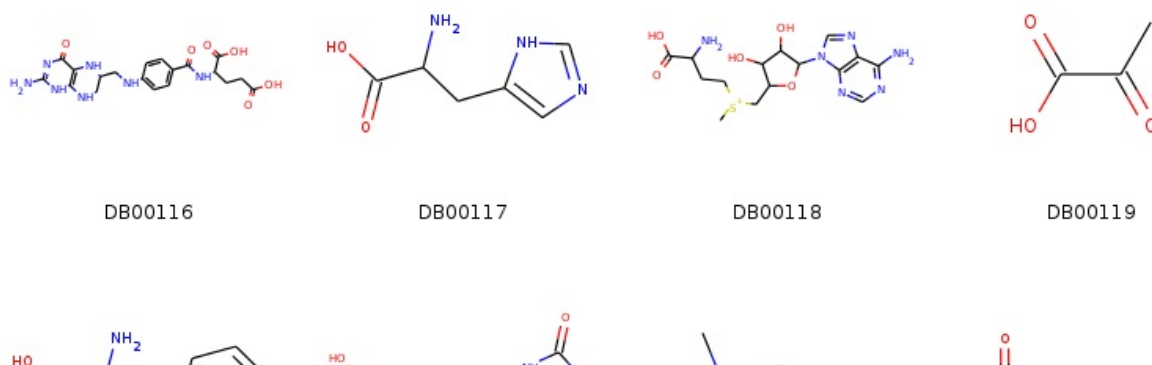
Out[36]:

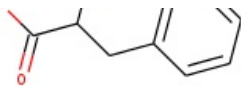
	DRUGBANK_ID	SMILES	ROMol	logp	mw
6	DB00116	<chem>Nc1nc(=O)c2c([nH]1)NCC(CNc1ccc(C(=O)NC(CCC(=O)O)C(=O)O)cc1)N2</chem>		-0.282	445.436

FrameToGridImage(pandasFrame, legendsCol=, molsPerRow=)

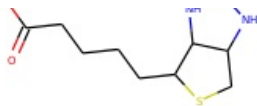
```
In [38]: PandasTools.FrameToGridImage(cpds.head(8), legendsCol=molnames, molsPerRow=4)
```

Out[38]:

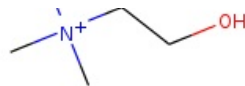




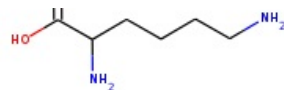
DB00120



DB00121



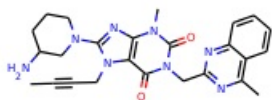
DB00122



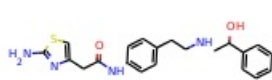
DB00123

In [39]: `PandasTools.FrameToGridImage(cpbs.tail(6), legendsCol='mw', molsPerRow=3)`

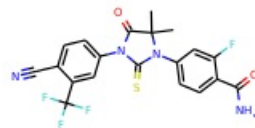
Out [39]:



472.553



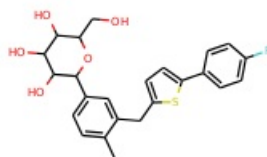
396.516



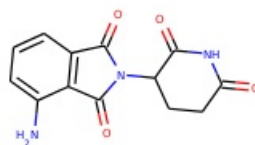
464.444



302.414



444.524



273.248

## Murcko scaffold decomposition

Removes side chain atoms

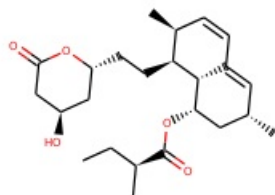
In [40]: `from rdkit.Chem.Scaffolds import MurckoScaffold`

How it works with RDKit:

In [41]: `scaffold = MurckoScaffold.GetScaffoldForMol(mol)  
generic = MurckoScaffold.MakeScaffoldGeneric(MurckoScaffold.GetScaffoldForMol(mol))`

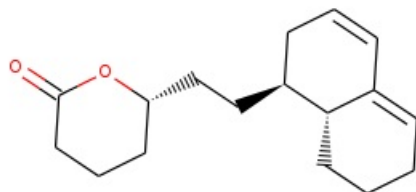
In [42]: `mol`

Out [42]:



In [43]: `scaffold`

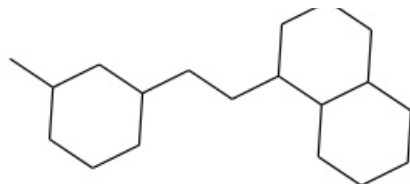
Out [43]:



In [44]: `generic`

Out [44]:





**AddMurckoToFrame(pandasFrame, MurckoCol=, Generic=False)**

Returns SMILES of scaffolds

```
In [45]: % time PandasTools.AddMurckoToFrame(cpds)
```

```
CPU times: user 403 ms, sys: 6.67 ms, total: 410 ms
Wall time: 408 ms
```

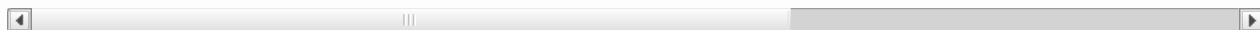
```
In [46]: % time PandasTools.AddMurckoToFrame(cpds, MurckoCol='Murcko_GENERIC', Generic=True)
```

```
CPU times: user 967 ms, sys: 10 ms, total: 977 ms
Wall time: 961 ms
```

```
In [47]: cpds.head(1)
```

Out [47]:

	DRUGBANK_ID	SMILES	ROMol	logp	mw	Murcko_
6	DB00116	Nc1nc(=O)c2c([nH]1)NCC(CNc1ccc(C(=O)NC(CCC(=O)O)C(=O)O)cc1)N2		-0.282	445.436	O=c1nc[



Now we can use pandas groupby() functionality and group by scaffolds and create a new frame with scaffolds sorted by number of members

```
In [48]: sortedScaffolds = cpds.groupby(['Murcko_SMILES']).count().sort(smiles, ascending=False)
```

```
In [49]: sortedScaffolds = sortedScaffolds[[smiles]] # Keep only smiles column
sortedScaffolds = sortedScaffolds.rename(columns={smiles:'count'}) # rename smiles column to count
sortedScaffolds['Murcko_SMILES'] = sortedScaffolds.index # actual SMILES are only in index column,
move it
sortedScaffolds.head()
```

Out [49]:


	count	Murcko_SMILES
Murcko_SMILES		
c1ccccc1	112	c1ccccc1
	111	
O=C1C=CC2C(=C1)CCC1C3CCCC3CCC21	17	O=C1C=CC2C(=C1)CCC1C3CCCC3CCC21
O=C1C=C2CCC3C4CCCC4CCC3C2CC1	12	O=C1C=C2CCC3C4CCCC4CCC3C2CC1
O=C1CN=C(c2ccccc2)c2ccccc2N1	12	O=C1CN=C(c2ccccc2)c2ccccc2N1

```
In [50]: PandasTools.AddMoleculeColumnToFrame(sortedScaffolds, smilesCol='Murcko_SMILES')
```

```
In [51]: sortedScaffolds.head(1)
```

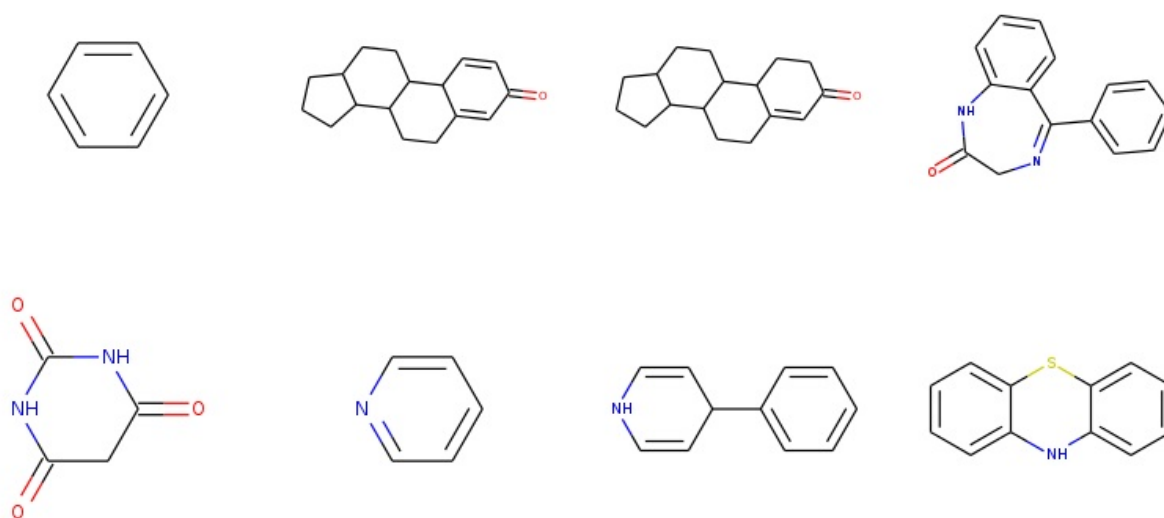
Out [51]:

	count	Murcko_SMILES	ROMol
Murcko_SMILES			

<chem>c1ccccc1</chem>	112	<chem>c1ccccc1</chem>	
-----------------------	-----	-----------------------	--

In [53]: `PandasTools.FrameToGridImage(sortedScaffolds.dropna().head(8), molsPerRow=4) #dropna drops compounds without scaffold`

Out [53]:




We can also retrieve all compounds with certain scaffold from original table

Benzodiazepine scaffold #4 O=C1CN=C(c2ccccc2)c2ccccc2N1

In [55]: `cpds[cpds['Murcko_SMILES'] == 'O=C1CN=C(c2ccccc2)c2ccccc2N1'].head(1)`

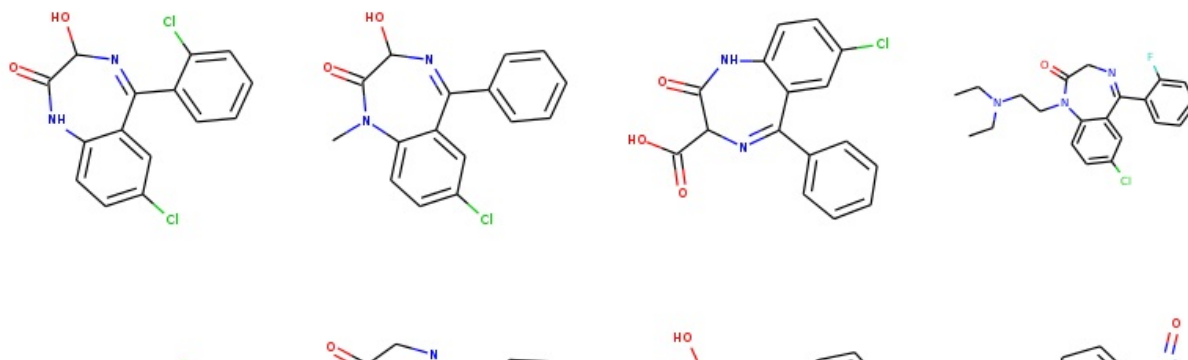
Out [55]:

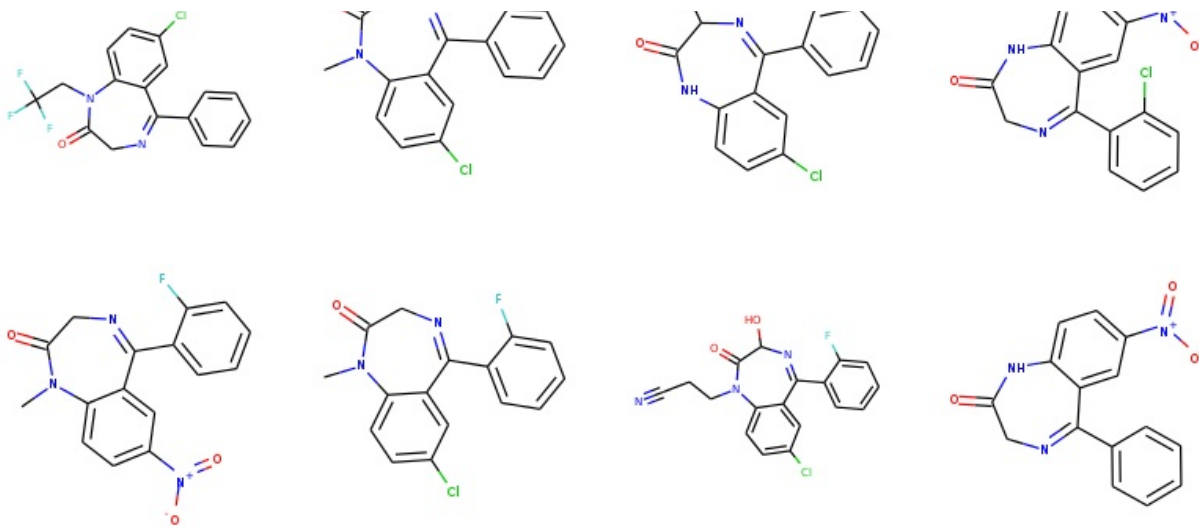
	DRUGBANK_ID	SMILES	ROMol	logp	mw	Murcko_SMILES	M
74	DB00186	<chem>O=C1Nc2ccc(Cl)cc2C(c2ccccc2Cl)=NC1O</chem>		3.1013	321.163	<chem>O=C1CN=C(c2ccccc2)c2ccccc2N1</chem>	C



In [57]: `PandasTools.FrameToGridImage(cpds[cpds['Murcko_SMILES'] == 'O=C1CN=C(c2ccccc2)c2ccccc2N1'], molsPerRow=4)`

Out [57]:





## Aligning compounds to scaffolds

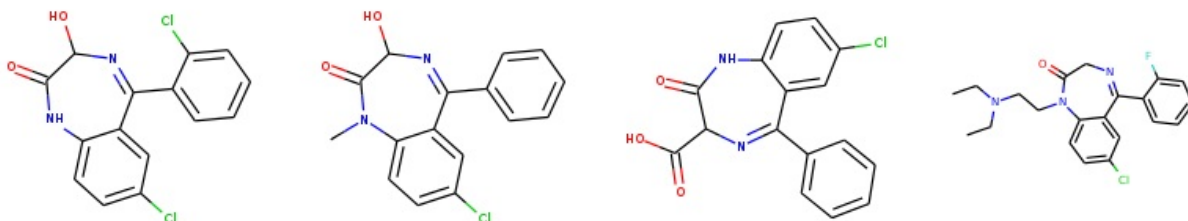
`AlignToScaffold(dataframe, molCol=, scaffoldCol=)`

```
In [58]: somemols = cpds.groupby('Murcko_SMILES').get_group('O=C1CN=C(c2ccccc2)c2ccccc2N1')
```

Note how molecules are not aligned

```
In [59]: PandasTools.FrameToGridImage(somemols.head(4), molsPerRow=4)
```

Out [59]:

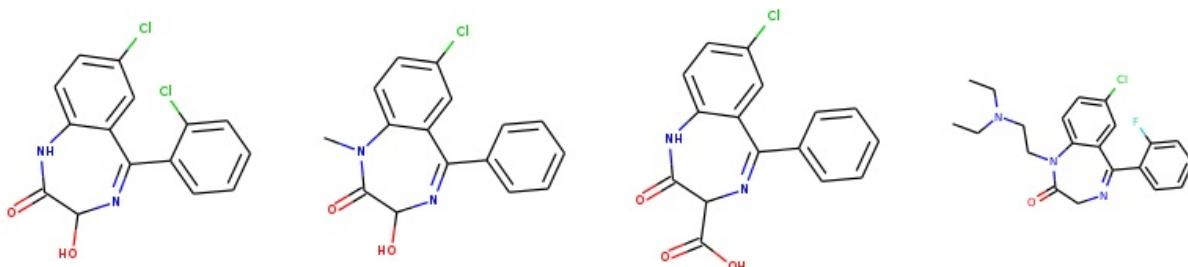


Align them to scaffold

```
In [60]: PandasTools.AlignToScaffold(somemols, molCol='ROMol', scaffoldCol='Murcko_SMILES')
```

```
In [61]: PandasTools.FrameToGridImage(somemols.head(4), molsPerRow=4)
```

Out [61]:



Copyright (C) 2013 by Samo Turk, [BioMed X GmbH \(http://bio.mx/\)](http://bio.mx/)

This work is licensed under the Creative Commons Attribution-ShareAlike 3.0 License. To view a copy of this license, visit <http://creativecommons.org/licenses/by-sa/3.0/> or send a letter to Creative Commons, 543 Howard Street, 5th Floor, San Francisco, California, 94105, USA.