PREDICTING BIODEGRADABILITY

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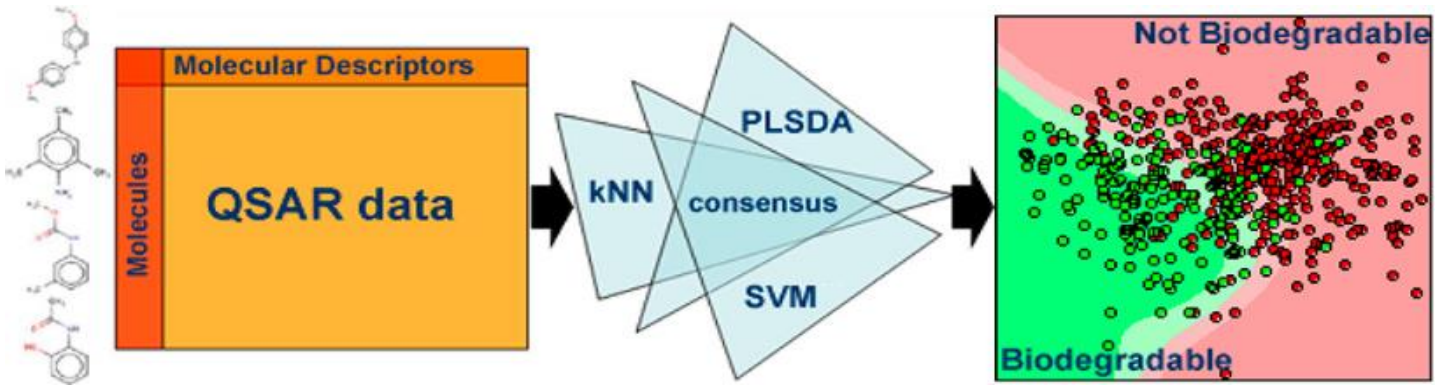
How to predict if a compound is biodegradable from the study of the relationships between chemical structure and biodegradation of molecules ?



SUMMARY

I - Data pre-processing

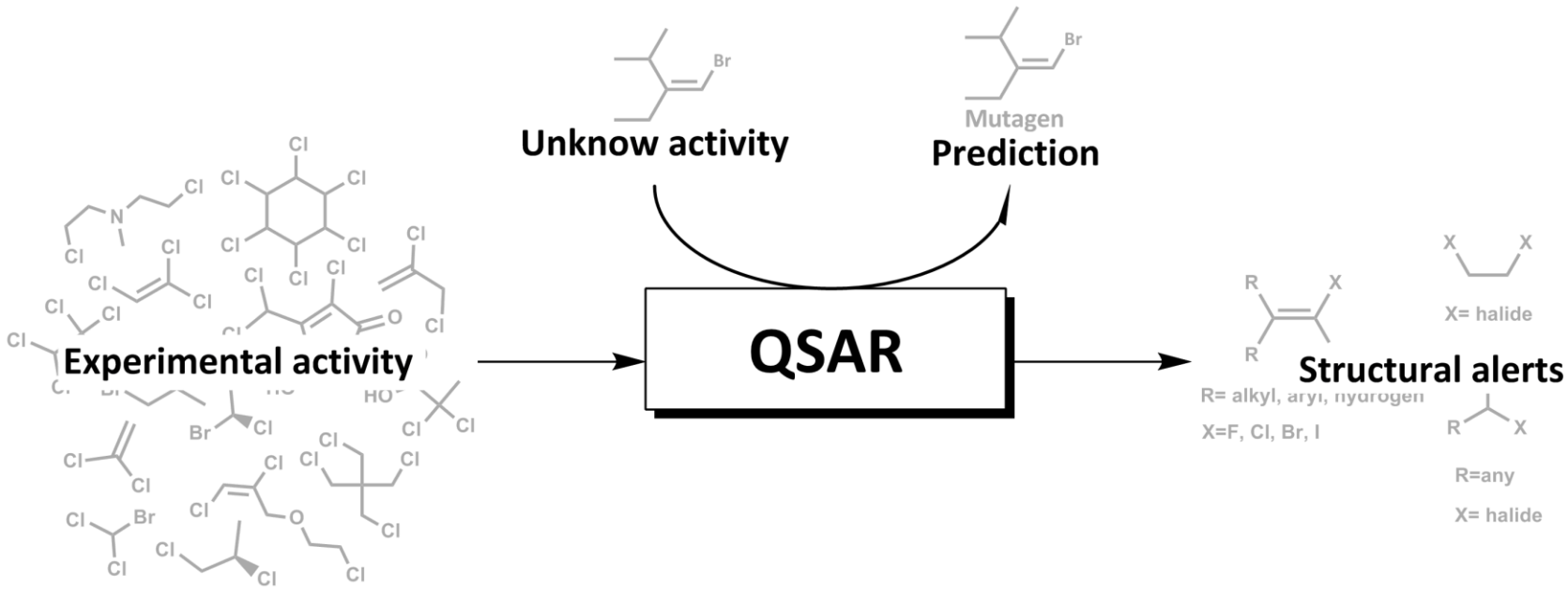
- **Import**
- **Cleaning**
- **Normalization**

II - Data visualization

- **Matplotlib**
- Correlation
- Seaborn

III – Modeling

- Random
- Logistic regression
- KNN
- Final model



THE DATA

Dataset used : QSAR biodegradation

Data set containing values for **41 attributes** (molecular descriptors) used to classify 1055 chemicals into **2 classes** (ready and not ready biodegradable).

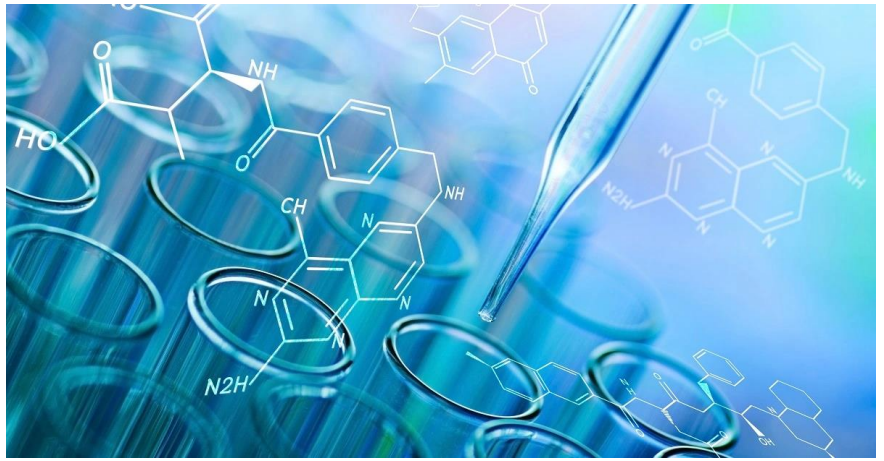
The data have been used to develop QSAR (**Quantitative Structure Activity Relationships**) models for the study of the relationships between chemical structure and biodegradation of molecules.

What the database allows :

- Binary classification
- Mathematical computation of properties

	SpMax_L	J_Dz(e)	nHM	F01[N-N]	F04[C-N]	NssssC	nCb-	C%	nCp	nO	...
0	3.919	2.6909	0	0	0	0	0	31.4	2	0	...
1	4.170	2.1144	0	0	0	0	0	30.8	1	1	...
2	3.932	3.2512	0	0	0	0	0	26.7	2	4	...
3	3.000	2.7098	0	0	0	0	0	20.0	0	2	...
4	4.236	3.3944	0	0	0	0	0	29.4	2	4	...
...	...	...	...	...	...	...	...	...	...	...	...
1050	5.431	2.8955	0	0	0	2	0	32.1	4	1	...
1051	5.287	3.3732	0	0	9	0	0	35.3	0	9	...
1052	4.869	1.7670	0	1	9	0	5	44.4	0	4	...
1053	5.158	1.6914	2	0	36	0	9	56.1	0	0	...
1054	5.076	2.6588	2	0	0	0	4	54.5	0	0	...

1055 rows × 42 columns



DATA PRE-PROCESSING

The DataFrame is missing column descriptions. These have to be added from the source website : [QSAR biodegradation - UCI Machine Learning Repository]

Only one object column : experimental class.
This is our target, we renamed it *classD*.

There is no missing value in the dataset.

Cleaning

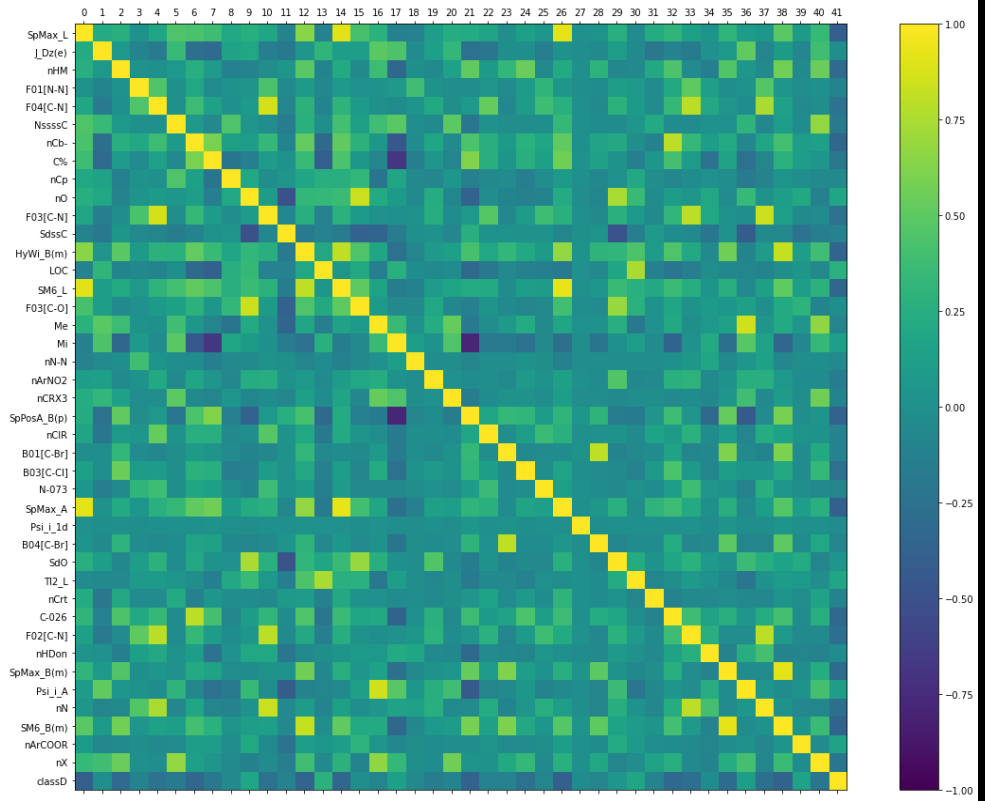
```
#look for types of values contains in the dataset
df.dtypes.value_counts()
```

```
int64      24
float64    17
object      1
dtype: int64
```

The target class will be expressed as integers :
NRB = 0 --> Not Biodegradable
RB = 1 --> Biodegradable

```
df.replace(["RB","NRB"],[1,0], inplace = True)
```

DATA VISUALIZATION



Heatmap of the correlation matrix for the columns in the df, providing a visual representation of how each column correlates with every other column.

```
0    0.662559
1    0.337441
```

→ NRB

→ RB

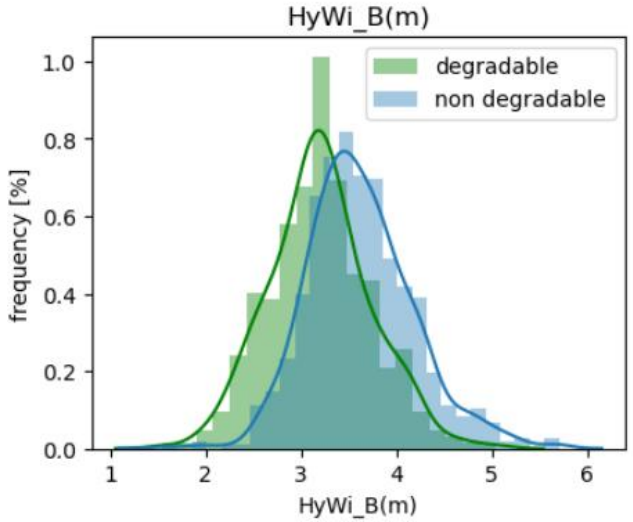
These percentages provide insights into the class distribution in the "degradable" column of the dataset => shows an imbalance.

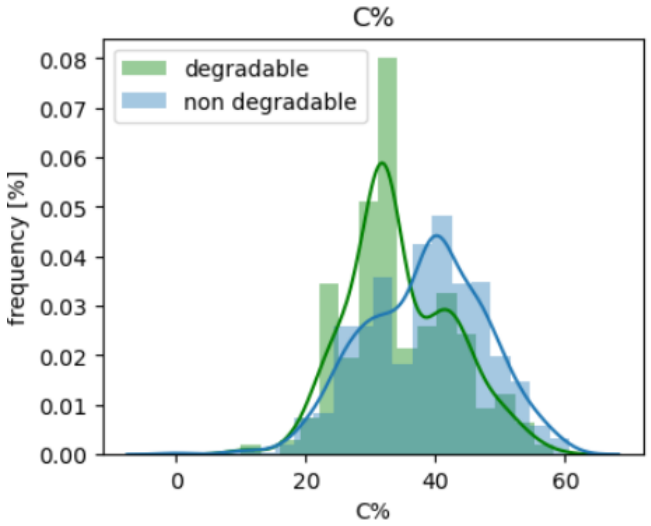
```
df.corr().applymap(lambda x: x if abs(x)>.90 else "")
```

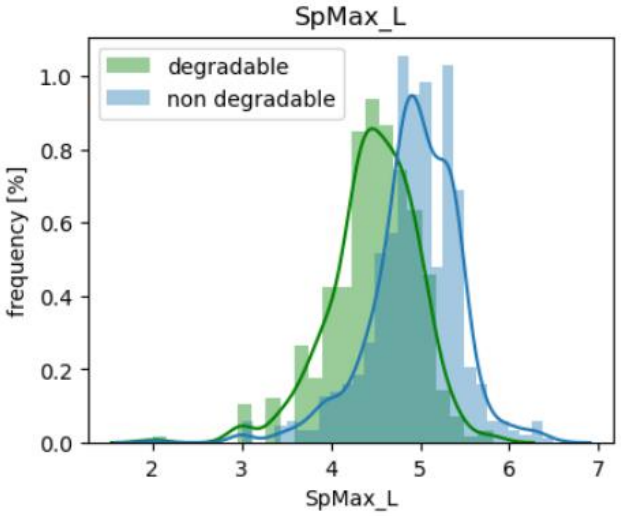
This operation is useful for highlighting strong correlations between variables while removing highly correlations (|r| > 0.90).
=> "SM6_L","SpMax_A","SM6_B(m)“ dropped

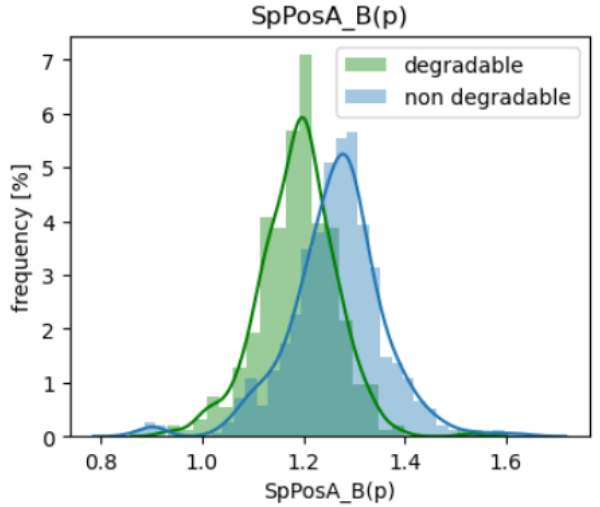
The distribution show that the best prediciting features for class seperation are:

- SpPosA_B(p)
- HyWi_B(m)
- C%
- SpMax_B(m)
- SpMax_L







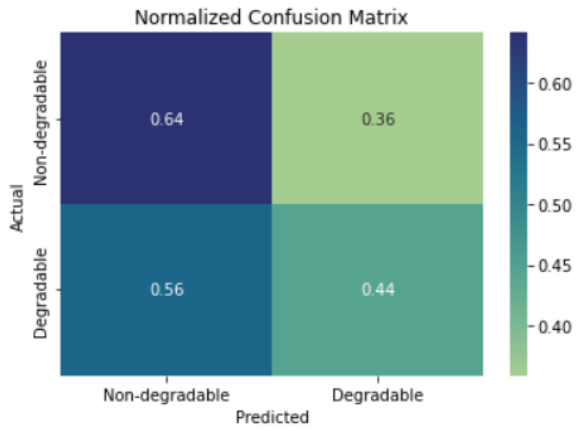


The plots shows how good the features are
able to divide the degradable and nondegradable substances.

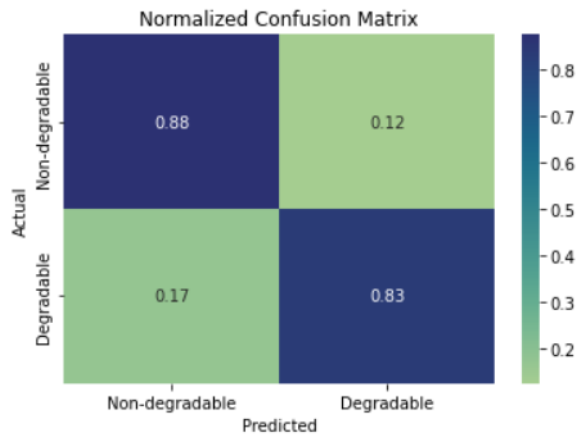
MODELING

In the context of modeling, we have decided to use different algorithms:

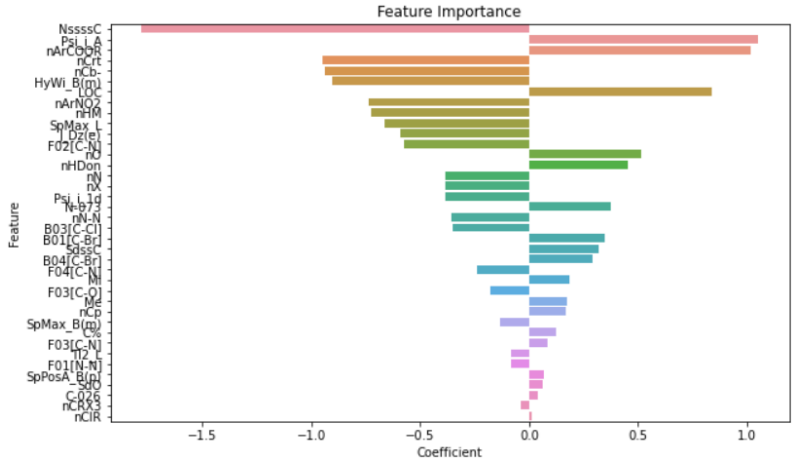
- Random
- Logistic Regression
- KNN (k-Nearest Neighbors)

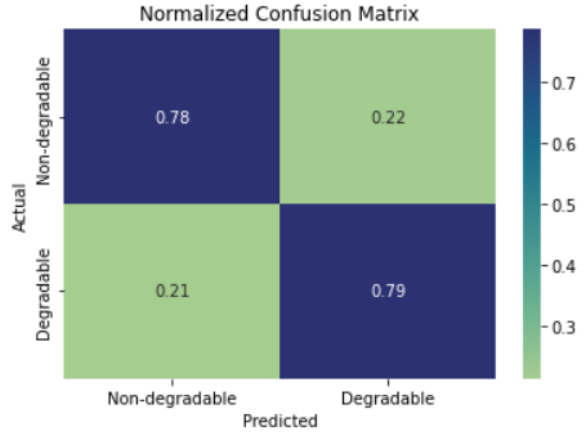


Random



Logistic regression





KNN

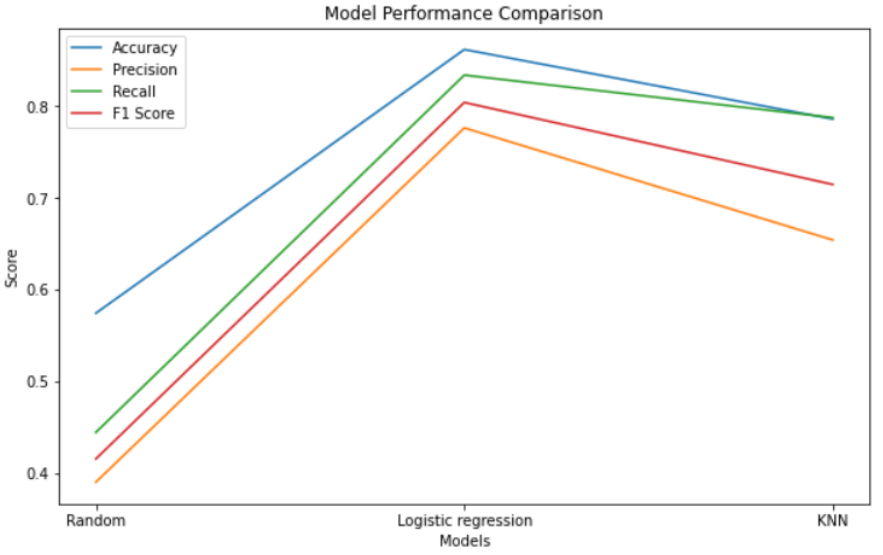
CONCLUSION

To explain the results of the above model classification, we need to analyze the performance metrics of each model.

The model with higher accuracy, precision, recall, and F1 score is considered better in terms of classification performance.

We can see that the best model tested for this classification problem is the Logistic regression model.

	model	accuracy	precision	recall	f1
0	Random	0.552050	0.336538	0.324074	0.330189
1	Logistic regression	0.861199	0.775862	0.833333	0.803571
2	KNN	0.785489	0.653846	0.787037	0.714286



THANK YOU FOR YOUR ATTENTION

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SOURCES

https://archive.ics.uci.edu/dataset/254/qsar+biodegradation

https://www.neuraldesigner.com/learning/examples/qsar-biodegradation/#DataSet

https://pubs.acs.org/doi/10.1021/ci4000213