
Supplementary Material

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1 Training details

2 For both experiments, toxicity estimation of PFAS molecules [1]¹ and biodegradability of general
3 molecules [2], we divided the datasets into: 60% for training purposes, 20% for validation purposes,
4 and 20% for testing purposes. In both cases, the divisions have been made following the original
5 papers in order to have a fair comparison with the state-of-the-art algorithms.

6 Table 1 refers to the hyper-parameters considered for the toxicity estimation task.

Table 1: MULTIMODAL-MOLFORMER hyper-parameters for toxicity prediction considering the LDToxDB dataset.

Hyper-parameter	Value
Device	cuda
GPU	1
Batch Size	32
Number of heads	12
Number of embeddings	768
Dropout	0.1
Learning rate	3e-5
Number of workers	8
Number of epochs	2000
Dimension of the Feed-Forward network	[797 797 797 1]

7 Table 2 refers to the hyper-parameters considered for biodegradability classification..

Table 2: MULTIMODAL-MOLFORMER hyper-parameters for biodegradability prediction.

Hyper-parameter	Value
Device	cuda
GPU	1
Batch Size	32
Number of heads	12
Number of layers	12
Number of embeddings	768
Dropout	0.1
Learning rate	3e-5
Number of workers	8
Number of epochs	2000
Dimension of the Feed-Forward network	[773 773 773 1]

8 Both experiments have been conducted in the IBM computing cluster with 32 GB of RAM memory
9 and a Nvidia V100 GPU.

10 The experiments conducted in this research are based on the recently introduced MOLFORMER
11 approach which can be accessed through this link (<https://github.com/IBM/molformer>).

¹AI4PFAS project <https://github.com/AI4PFAS/AI4PFAS>.

12 To extract the physico-chemical features based on the Mordred descriptors we used the library avail-
13 able at [http://mordred-descriptor.github.io/documentation/v0.1.0/introduction.](http://mordred-descriptor.github.io/documentation/v0.1.0/introduction.html)
14 [html](http://mordred-descriptor.github.io/documentation/v0.1.0/introduction.html) [3]. To build the Markov-Blanket causal feature selection blocks we used the library available
15 at <https://github.com/atif-hassan/PyImpetus> [4]. The default parameters were used for
16 the physico-chemical extraction from SMILES and to build the Markov-Blanket causal graphs.

17 **References**

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