

Supporting Information

Transformer-based deep learning method for optimizing ADMET properties of lead compounds

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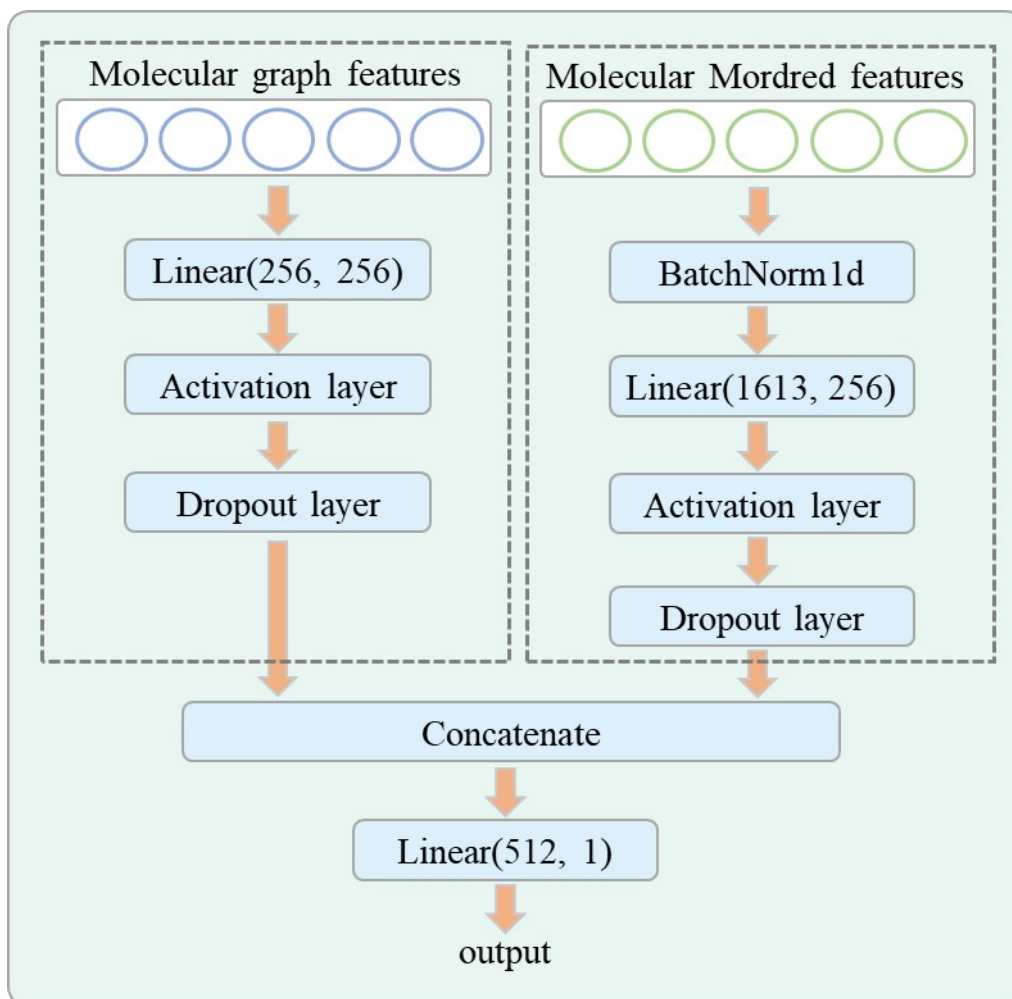


Figure S1. The detailed flow of the prediction layer.

Table S1. The hyperparameter settings for predictive and optimization models.

	ADMET prediction model	Constraints-Transformer
Parameters	Optimizer=Adam	Optimizer=Adam
	Loss Function=BCE/ SmoothL1	Loss Function=KLDivLoss
	Lr=1e-5	Num_encoder_layer=6
	Num_encoder_layer=3	Num_decoder_layer=6
	Num_heads=8	Num_heads=8
	D_model=256	D_model=256
	D_ff = 1024	D_ff = 2048
	Dense_dropout=0.5	Epoch=100
	Dense_activation =LeakyRELU(0.1)	Batch_size=256
	Epoch=200	Num_samples=10
	Batch_size=24/32	

Table S2. The label of each property.

Property	Measured unit	Threshold	Threshold in <i>ln</i> scale	Discrete property labels	Changed property labels
<i>lnS</i>	mol/L	Low: ≤ 0.05 High: > 0.05	Low: ≤ -3.0 High: > -3.0	$\ln S_{(\text{inf}, -7)}$	$\ln S_{\text{low-high}}$
				$\ln S_{-7}$	$\ln S_{\text{high-low}}$
				$\ln S_{-6}$	$\ln S_{\text{no-change}}$
				...	
				$\ln S_{-2}$	
				$\ln S_{(2, \text{inf})}$	
<i>logD</i>	--	--	--	$\log D_{(\text{inf}, -5)}$	$\log D_{(-\text{inf}, -5]}$
				$\log D_{-5}$	$\log D_{(-5, -4]}$
				$\log D_{-4}$	$\log D_{(-4, -3]}$
			
				$\log D_{-5}$	$\log D_{(4, 5]}$
				$\log D_{(5, \text{inf})}$	$\log D_{(5, \text{inf})}$
<i>CL</i>	$\mu\text{L}/\text{min}/\text{mg}$	Low: ≤ 20 High: > 20	Low: ≤ 3.0 High: > 3.0	$CL_{(\text{inf}, -5)}$	$CL_{\text{low-high}}$
				CL_{-5}	$CL_{\text{high-low}}$
				CL_{-4}	$CL_{\text{no-change}}$
				...	
				CL_{-5}	
				$CL_{(5, \text{inf})}$	

Table S3. Evaluation result on TDC ADMET benchmark group.

Dataset	Matrix	Current Top1	Bert & Mordred	Only Bert	Only Mordred	Rank
caco2	MAE^{\downarrow}	0.288±0.011	0.277±0.002	0.331±0.004	0.321±0.010	1st
lipophilicity		0.533±0.005	0.524±0.005	0.562±0.022	0.591±0.056	1st
solubility_aqsolddb		0.727±0.004	0.727±0.004	0.771±0.009	0.780±0.024	1st
ppbr		8.251±0.115	7.879±0.102	7.722±0.165	10.889±0.269	1st
ld50		0.588±0.005	0.603±0.003	0.607±0.165	0.656±0.009	3rd
hia	$AUROC^{\uparrow}$	0.988±0.002	0.979±0.008	0.936±0.021	0.933±0.020	3rd
pgp		0.946±0.001	0.936±0.000	0.889±0.010	0.931±0.004	2nd
bioav		0.748±0.033	0.685±0.010	0.668±0.037	0.721±0.005	3rd
bbb		0.905±0.001	0.911±0.002	0.868±0.007	0.899±0.004	1st
cyp3a4 substrate		0.680±0.005	0.681±0.000	0.652±0.008	0.669±0.012	1st
herg		0.874±0.014	0.835±0.014	0.789±0.017	0.801±0.013	3rd
Ames		0.865±0.002	0.855±0.001	0.804±0.006	0.844±0.002	3rd
dili		0.933±0.011	0.933±0.005	0.903±0.003	0.908±0.005	1st
cyp2c9 inhibition		$AUPRC^{\uparrow}$	0.794±0.004	0.745±0.002	0.709±0.006	0.737±0.005
cyp2d6 inhibition	0.721±0.001		0.624±0.006	0.655±0.009	0.611±0.013	3rd
cyp3a4 inhibition	0.882±0.001		0.849±0.003	0.803±0.011	0.852±0.003	3rd
cyp2c9 substrate	0.433±0.017		0.444±0.014	0.396±0.015	0.416±0.022	1st
cyp2d6 substrate	0.677±0.047		0.725±0.012	0.579±0.006	0.678±0.030	1st
Vdss	ρ^{\uparrow}	0.612±0.018	0.632±0.015	0.585±0.014	0.583±0.026	1st
Half life		0.416±0.009	0.485±0.020	0.396±0.021	0.492±0.029	1st
Cl-hepa		0.491±0.006	0.422±0.011	0.364±0.031	0.405±0.008	3rd
Cl-micro		0.625±0.002	0.583±0.005	0.549±0.015	0.578±0.013	6th