gNetDGP: An End-to-End Graph Neural Network for Disease Gene Prioritization

Supplementary Material

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Table 1: AUC-ROC performances of SmuDGE and gNetDGP on monogenic and multigenic diseases using generic setting. Best performances are in bold.

Disease Type	\mathbf{SmuDGE}	${f gNetDGP}$
Monogenic	0.80	0.86
Multigenic	0.89	0.94

Table 2: Experimental performances of DiGI, SmuDGE and gNetDGP on 12 disease classes. For each metric, the highest value is in bold face.

Disease Class	Metric	DiGI	SmuDGE	gNetDGP
Ophthamological	auc-roc	0.71	0.80	0,86
	auc-pr	0.71	0.83	0,87
Connective	auc-roc	0.71	0.43	0,72
	auc-pr	0.75	0.45	0,76
Endocrine	auc-roc	0.76	0.82	0,82
	auc-pr	0.78	0.82	0,84
Skeletal	auc-roc	0.81	0.65	0,89
	auc-pr	0.84	0.70	$0,\!90$
Metabolic	auc-roc	0.83	0.85	0,89
	auc-pr	0.87	0.81	0,89
Cardiovascular	auc-roc	0.71	0.69	0,74
	auc-pr	0.77	0.67	$0,\!77$
Dermatological	auc-roc	0.79	0.82	0,88
	auc-pr	0.85	0.80	$0,\!91$
Renal	auc-roc	0.72	0.76	0,73
	auc-pr	0.72	0.77	0,77
Hematological	auc-roc	0.70	0.68	0,86
	auc-pr	0.75	0.70	0,88
Immunological	auc-roc	0.86	0.82	0,89
	auc-pr	0.92	0.76	0,91
Muscular	auc-roc	0.74	0.78	0,84
	auc-pr	0.81	0.82	0,86
Developmental	auc-roc	0.69	0.65	0,70
	auc-pr	0.64	0.72	0,69