

# Supplemental Material for Label-Specific Multi-Semantics Metric Learning for Multi-Label Classification: Global Consideration Helps

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## 1 Further Experimental Results

### 1.1 Comparative Studies

We employ ten-fold cross-validation to evaluate the proposed LIMIC approach and three well-established multi-label metric learning approaches  $\mathcal{B} \in \{\text{LM}, \text{LJE}, \text{COMMU}\}$  on 8 benchmark data sets. Table 7 reports detailed experimental results in terms of *Average precision*, *Macro-F1*, and *Macro-averaging AUC*, which are not covered in the ‘‘Comparative Studies’’ subsection of the main body due to page limit. Furthermore, pairwise *t*-test at 0.05 significance level is conducted to demonstrate whether the performance difference between  $\mathcal{A}$ -LIMIC and  $\mathcal{A}$ - $\mathcal{B}$  is significant, where the resulting win/tie/loss counts are reported in Table 1, 2, and 3.

### 1.2 Further Analyses

Pairwise *t*-test at 0.05 significance level is conducted to analyze whether LIMIC performs statistically better than its variants. Tabel 4, 5, and 6 summarize win/tie/loss counts on each evaluation metric, which demonstrate LIMIC is statistically superior to its variants in terms of all evaluation metrics.

Metrics	BR-KNN-LIMIC against			
	BR-KNN	BR-KNN-LM	BR-KNN-LJE	BR-KNN-COMMU
<i>Hamming Loss</i>	7/1/0	5/1/2	7/1/0	5/3/0
<i>Ranking Loss</i>	6/2/0	5/3/0	6/2/0	2/6/0
<i>Coverage</i>	6/2/0	6/2/0	6/2/0	4/4/0
<i>Average precision</i>	6/2/0	8/0/0	6/2/0	4/4/0
<i>Macro-F1</i>	7/1/0	4/3/1	7/1/0	5/3/0
<i>Macro-averaging AUC</i>	6/2/0	6/2/0	6/2/0	3/5/0
<b>In Total</b>	<b>38/10/0</b>	<b>34/11/3</b>	<b>38/10/0</b>	<b>23/25/0</b>

Table 1: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for BR-KNN-LIMIC against other compared approaches.

Metrics	ML-KNN-LIMIC against			
	ML-KNN	ML-KNN-LM	ML-KNN-LJE	ML-KNN-COMMU
<i>Hamming Loss</i>	5/3/0	4/3/1	5/3/0	4/4/0
<i>Ranking Loss</i>	7/1/0	5/3/0	7/1/0	4/4/0
<i>Coverage</i>	5/3/0	4/4/0	5/3/0	3/4/1
<i>Average precision</i>	6/2/0	6/2/0	6/2/0	5/3/0
<i>Macro-F1</i>	6/2/0	6/2/0	6/2/0	6/2/0
<i>Macro-averaging AUC</i>	6/2/0	5/3/0	6/2/0	4/4/0
<b>In Total</b>	<b>35/13/0</b>	<b>30/17/1</b>	<b>35/13/0</b>	<b>26/21/1</b>

Table 2: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for ML-KNN-LIMIC against other compared approaches.

\*The work was done when Wei Wang was with Southeast University.

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Metrics	RELIAB-KNN-LIMIC against			
	RELIAB-KNN	RELIAB-KNN-LM	RELIAB-KNN-LJE	RELIAB-KNN-COMMU
<i>Hamming Loss</i>	6/2/0	6/2/0	6/1/1	6/2/0
<i>Ranking Loss</i>	6/2/0	5/3/0	6/2/0	5/3/0
<i>Coverage</i>	7/1/0	6/0/2	6/1/1	7/1/0
<i>Average precision</i>	5/3/0	7/0/1	7/1/0	6/1/1
<i>Macro-F1</i>	8/0/0	8/0/0	8/0/0	8/0/0
<i>Macro-averaging AUC</i>	7/0/1	8/0/0	8/0/0	6/0/2

Table 3: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for RELIAB-KNN-LIMIC against other compared approaches.

Metrics	BR-KNN-LIMIC against		
	BR-KNN-LIMIC-NG	BR-KNN-LIMIC-NL	BR-KNN-LIMIC-NGL
<i>Hamming Loss</i>	4/4/0	3/4/1	4/3/1
<i>Ranking Loss</i>	7/1/0	5/3/0	7/1/0
<i>Coverage</i>	6/1/1	4/3/1	5/2/1
<i>Average precision</i>	7/1/0	6/2/0	8/0/0
<i>Macro-F1</i>	7/0/1	3/4/1	6/1/1
<i>Macro-averaging AUC</i>	6/1/1	4/2/2	8/0/0
<b>In Total</b>	<b>37/8/3</b>	<b>25/18/5</b>	<b>38/7/3</b>

Table 4: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for BR-KNN-LIMIC against its variants coupled with BR-KNN.

Metrics	ML-KNN-LIMIC against		
	ML-KNN-LIMIC-NG	ML-KNN-LIMIC-NL	ML-KNN-LIMIC-NGL
<i>Hamming Loss</i>	6/2/0	4/4/0	4/4/0
<i>Ranking Loss</i>	5/2/1	5/3/0	5/2/1
<i>Coverage</i>	4/2/2	3/3/2	4/3/1
<i>Average precision</i>	6/1/1	5/1/2	6/1/1
<i>Macro-F1</i>	7/0/1	3/3/2	5/1/2
<i>Macro-averaging AUC</i>	6/0/2	5/3/0	6/1/1
<b>In Total</b>	<b>34/7/7</b>	<b>25/17/6</b>	<b>30/13/6</b>

Table 5: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for ML-KNN-LIMIC against its variants coupled with ML-KNN.

Metrics	RELIAB-KNN-LIMIC against		
	RELIAB-KNN-LIMIC-NG	RELIAB-KNN-LIMIC-NL	RELIAB-KNN-LIMIC-NGL
<i>Hamming Loss</i>	5/3/0	6/2/0	5/3/0
<i>Ranking Loss</i>	4/2/2	5/3/0	6/1/1
<i>Coverage</i>	7/1/0	4/3/1	5/3/0
<i>Average precision</i>	6/1/1	4/3/1	4/4/0
<i>Macro-F1</i>	6/1/1	7/0/1	5/2/1
<i>Macro-averaging AUC</i>	7/0/1	5/3/0	7/1/0
<b>In Total</b>	<b>35/8/5</b>	<b>31/14/3</b>	<b>32/14/2</b>

Table 6: Win/tie/loss counts (pairwise *t*-test at 0.05 significant level) for RELIAB-KNN-LIMIC against its variants coupled with RELIAB-KNN.

Compared Algorithms	Data Sets							
	CAL500	emotions	birds	genbase	medical	image	scene	yeast
	Average precision $\uparrow$							
BR-KNN	0.463±0.009	0.700±0.049●	0.358±0.043●	0.993±0.005	0.801±0.025●	0.788±0.023●	0.850±0.012●	0.762±0.018●
BR-KNN-LM	0.451±0.007●	0.711±0.038●	0.332±0.052●	0.989±0.009●	0.843±0.032●	0.789±0.020●	0.847±0.013●	0.744±0.021●
BR-KNN-LJE	0.456±0.009	0.771±0.042●	0.390±0.042●	0.994±0.005	0.767±0.030●	0.767±0.024●	0.812±0.017●	0.750±0.019●
BR-KNN-COMMU	<b>0.467±0.010</b>	0.700±0.049●	0.356±0.043●	0.993±0.005	0.802±0.025●	0.789±0.023●	0.850±0.012	0.763±0.018
BR-KNN-LIMIC	0.464±0.010	<b>0.800±0.040</b>	<b>0.441±0.065</b>	<b>0.996±0.005</b>	<b>0.881±0.024</b>	<b>0.807±0.024</b>	<b>0.857±0.019</b>	<b>0.767±0.022</b>
ML-KNN	0.494±0.008	0.712±0.042●	0.414±0.052●	0.989±0.008	0.819±0.020●	0.789±0.021●	0.867±0.017●	0.765±0.018●
ML-KNN-LM	0.493±0.007	0.719±0.019●	0.405±0.056●	0.995±0.006	0.849±0.035●	0.789±0.017●	0.857±0.015●	0.753±0.025●
ML-KNN-LJE	0.491±0.007	0.762±0.036●	0.425±0.039●	0.995±0.004	0.773±0.025●	0.772±0.023●	0.817±0.020●	0.752±0.018●
ML-KNN-COMMU	<b>0.497±0.010</b>	0.712±0.042●	0.413±0.052●	0.989±0.008●	0.817±0.020●	0.788±0.021●	0.867±0.016	0.764±0.019
ML-KNN-LIMIC	0.495±0.008	<b>0.789±0.037</b>	<b>0.461±0.054</b>	<b>0.996±0.002</b>	<b>0.891±0.019</b>	<b>0.811±0.025</b>	<b>0.868±0.013</b>	<b>0.768±0.022</b>
RELIAB-KNN	0.589±0.010	0.797±0.040●	0.498±0.035●	0.995±0.007	0.857±0.023●	0.831±0.025	0.885±0.015●	0.793±0.020●
RELIAB-KNN-LM	0.583±0.008●	0.816±0.038●	0.531±0.033●	0.987±0.006●	0.873±0.024●	<b>0.849±0.027</b> ○	0.904±0.013●	0.807±0.024●
RELIAB-KNN-LJE	0.588±0.010	0.835±0.035●	0.524±0.036●	0.984±0.005●	0.826±0.025●	0.816±0.023●	0.852±0.016●	0.784±0.018●
RELIAB-KNN-COMMU	<b>0.591±0.009</b> ○	0.796±0.040●	0.498±0.035●	<b>0.996±0.007●</b>	0.857±0.023●	0.831±0.025	0.885±0.015●	0.794±0.020●
RELIAB-KNN-LIMIC	0.589±0.012	<b>0.857±0.037</b>	<b>0.574±0.041</b>	<b>0.996±0.008</b>	<b>0.898±0.028</b>	0.830±0.026	<b>0.927±0.018</b>	<b>0.813±0.024</b>
Macro-F1 $\uparrow$								
BR-KNN	0.095±0.005●	0.463±0.045	0.038±0.018●	0.583±0.039●	0.169±0.018●	0.591±0.034●	0.721±0.014●	0.416±0.019●
BR-KNN-LM	<b>0.103±0.006</b> ○	0.490±0.033●	0.060±0.035●	0.645±0.057	0.321±0.051●	0.609±0.033	0.738±0.026	0.389±0.023●
BR-KNN-LJE	0.089±0.005●	0.579±0.049	0.045±0.037●	0.636±0.050●	0.219±0.033●	0.549±0.033●	0.665±0.027●	0.383±0.015●
BR-KNN-COMMU	0.097±0.007	0.463±0.045●	0.038±0.018●	0.583±0.039●	0.170±0.018●	0.591±0.034●	0.721±0.014	0.417±0.021
BR-KNN-LIMIC	0.097±0.006	<b>0.615±0.016</b>	<b>0.103±0.028</b>	<b>0.656±0.042</b>	<b>0.398±0.043</b>	<b>0.621±0.033</b>	<b>0.743±0.031</b>	<b>0.423±0.027</b>
ML-KNN	0.053±0.003●	0.361±0.050	0.010±0.007●	0.525±0.034	0.210±0.027●	0.575±0.030●	0.737±0.021●	0.367±0.011●
ML-KNN-LM	0.053±0.003	0.438±0.036●	0.016±0.011●	0.561±0.036●	0.295±0.043●	0.584±0.025●	0.735±0.026	0.342±0.018●
ML-KNN-LJE	0.048±0.002●	0.544±0.043	0.025±0.024●	0.598±0.046	0.214±0.017●	0.521±0.026●	0.656±0.026●	0.332±0.012●
ML-KNN-COMMU	<b>0.055±0.003</b>	0.361±0.050●	0.010±0.007●	0.525±0.034●	0.212±0.027●	0.576±0.031●	0.738±0.022	0.367±0.015●
ML-KNN-LIMIC	0.053±0.003	<b>0.593±0.033</b>	<b>0.072±0.030</b>	<b>0.609±0.034</b>	<b>0.371±0.024</b>	<b>0.643±0.036</b>	<b>0.747±0.019</b>	<b>0.419±0.024</b>
RELIAB-KNN	0.135±0.004●	0.512±0.034●	0.164±0.019●	0.612±0.036●	0.289±0.022●	0.634±0.032●	0.745±0.015●	0.458±0.012●
RELIAB-KNN-LM	0.147±0.002●	0.577±0.029●	0.194±0.016●	0.679±0.033●	0.358±0.018●	0.629±0.034●	0.752±0.019●	0.436±0.018●
RELIAB-KNN-LJE	0.129±0.003●	0.586±0.042●	0.189±0.022●	0.656±0.032●	0.347±0.020●	0.598±0.027●	0.710±0.026●	0.427±0.015●
RELIAB-KNN-COMMU	0.137±0.003●	0.512±0.034●	0.164±0.019●	0.612±0.036●	0.289±0.022●	0.633±0.032●	0.746±0.015●	0.458±0.012●
RELIAB-KNN-LIMIC	<b>0.155±0.003</b>	<b>0.623±0.033</b>	<b>0.234±0.020</b>	<b>0.726±0.042</b>	<b>0.395±0.031</b>	<b>0.682±0.035</b>	<b>0.758±0.016</b>	<b>0.482±0.016</b>
Macro-averaging AUC $\uparrow$								
BR-KNN	0.540±0.016	0.737±0.041●	0.670±0.040●	0.985±0.019	0.880±0.032●	0.840±0.019●	<b>0.936±0.006●</b>	<b>0.695±0.021●</b>
BR-KNN-LM	0.541±0.014	0.742±0.024●	0.658±0.056●	0.973±0.020●	0.879±0.051	0.826±0.020●	0.922±0.009●	0.655±0.030●
BR-KNN-LJE	0.525±0.018	0.807±0.030●	0.678±0.039●	0.986±0.015	0.837±0.033●	0.818±0.018●	0.905±0.012●	0.659±0.014●
BR-KNN-COMMU	<b>0.542±0.018</b>	0.737±0.041●	0.670±0.040●	0.986±0.019	0.880±0.031	0.840±0.019●	<b>0.936±0.006</b>	<b>0.695±0.022</b>
BR-KNN-LIMIC	<b>0.542±0.017</b>	<b>0.826±0.019</b>	<b>0.752±0.031</b>	<b>0.994±0.012</b>	<b>0.892±0.038</b>	<b>0.852±0.020</b>	<b>0.930±0.008</b>	<b>0.694±0.020</b>
ML-KNN	0.526±0.011	0.720±0.041●	0.666±0.044●	0.985±0.018	0.876±0.032●	0.834±0.019●	<b>0.933±0.006●</b>	0.677±0.020●
ML-KNN-LM	0.524±0.016	0.729±0.025●	0.642±0.062●	0.988±0.013	0.878±0.050	0.827±0.021●	0.921±0.009●	0.650±0.034●
ML-KNN-LJE	0.512±0.018	0.787±0.024●	0.664±0.048●	0.986±0.015	0.833±0.034●	0.811±0.018●	0.903±0.013●	0.652±0.014●
ML-KNN-COMMU	<b>0.530±0.011</b>	0.720±0.041●	0.664±0.044●	0.985±0.018●	0.875±0.031	0.834±0.019●	<b>0.933±0.006</b>	0.679±0.021
ML-KNN-LIMIC	0.525±0.011	<b>0.818±0.022</b>	<b>0.743±0.032</b>	<b>0.994±0.012</b>	<b>0.892±0.037</b>	<b>0.847±0.021</b>	0.929±0.009	<b>0.691±0.019</b>
RELIAB-KNN	0.651±0.013●	0.844±0.033●	0.672±0.042●	0.986±0.019●	0.897±0.035●	0.852±0.016●	<b>0.945±0.005</b> ○	0.734±0.020●
RELIAB-KNN-LM	0.653±0.010●	0.848±0.028●	0.643±0.035●	0.989±0.020●	0.898±0.029●	0.837±0.013●	0.934±0.004●	0.695±0.018●
RELIAB-KNN-LJE	0.633±0.011●	0.857±0.026●	0.667±0.038●	0.987±0.018●	0.845±0.027●	0.817±0.015●	0.928±0.003●	0.703±0.015●
RELIAB-KNN-COMMU	<b>0.659±0.012</b> ○	0.844±0.033●	0.672±0.042●	0.985±0.019●	0.897±0.035●	0.852±0.016●	<b>0.945±0.005</b> ○	0.734±0.020●
RELIAB-KNN-LIMIC	0.657±0.015	<b>0.896±0.030</b>	<b>0.721±0.039</b>	<b>0.995±0.018</b>	<b>0.913±0.033</b>	<b>0.878±0.020</b>	0.941±0.005	<b>0.786±0.016</b>

Table 7: Predictive performance of each compared approach (mean±std) in terms of *Average precision*, *Macro-F1*, and *Macro-average AUC*.  $\uparrow$  ( $\downarrow$ ) indicates the larger (smaller) the value, the better the performance. The best results are highlighted in **boldface**. In addition, ●/○ indicates whether  $\mathcal{A}$ -LIMIC ( $\mathcal{A} \in \{\text{BR-KNN, ML-KNN, RELIAB-KNN}\}$ ) achieves significantly superior/inferior to the compared approach on each data set in terms of different evaluation metrics (pairwise t-test at 0.05 significance level).