

# Ranking-based Convolutional Neural Network Models for Peptide-MHC Class I Binding Prediction (Supplementary Materials)

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## APPENDIX

### A1 ADDITIONAL EXPERIMENTAL RESULTS

#### A1.1 Hyperparameter Search Space

The search space of hyperparameters includes batch size {32, 128}, number of layers {1, 2}, filter size {1, 3, 5}, number of filters {8, 16, 64} and the size of output layer {8, 16, 64}.

#### A1.2 Model Architecture Comparison

Table A1, A2, A3 and A4 present the average percentage improvement of our models over the baseline MHCflurry with MS. Models are trained with BLOSUM +Onehot +Deep and selected with respect to 4 different metrics (i.e., AR<sub>100</sub>, HR<sub>100</sub>, AUC and ROC<sub>5</sub>, respectively).

#### A1.3 Encoding Method Comparison

Table A5, A6, A7, and A8 present the average percentage improvement of the 7 encoding methods over the baseline encoding method BLOSUM. We used SpConvM with H<sub>v</sub> as the model to evaluate encoding methods and their combination. We trained the models and selected the hyper-parameters with respect to 4 different metrics (i.e., AR<sub>100</sub>, HR<sub>100</sub>, AUC and ROC<sub>5</sub>, respectively).

**Table A1.** Performance Comparison over MHCflurry with MS (AR<sub>100</sub>; BLOSUM+Onehot+Deep)

Model	loss	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
ConvM	H <sub>v</sub>	8.08	4.79	3.56	5.93	6.36	9.92	10.01
	H <sub>l</sub>	6.36	4.50	1.33	4.50	5.21	6.57	5.92
	H <sub>i</sub>	6.58	3.49	1.36	3.26	5.63	7.25	6.05
	MS	-5.17	-7.93	-6.69	-1.93	1.24	-3.51	-3.36
SpConvM	H <sub>v</sub>	10.75	7.58	<b>8.06</b>	<b>9.61</b>	<b>8.68</b>	<b>18.84</b>	<b>16.29</b>
	H <sub>l</sub>	10.10	5.39	5.87	6.69	7.39	14.09	10.97
	H <sub>i</sub>	10.59	<b>8.28</b>	5.32	7.47	7.35	14.66	11.56
	MS	7.36	7.54	3.27	5.74	5.16	15.66	11.68
MHCflurry	H <sub>v</sub>	<b>10.95</b>	6.74	6.62	7.95	6.45	15.87	11.45
	H <sub>l</sub>	8.61	7.32	5.37	6.08	5.71	10.75	9.39
	H <sub>i</sub>	8.20	4.19	4.87	5.87	4.92	11.79	7.99
	MS	0.00	0.00	0.00	0.00	0.00	0.00	0.00

The values in the table are percentage improvement compared with the baseline MHCflurry with MS. Models are trained using BLOSUM+Onehot+Deep encoding methods, and selected with respect to AR<sub>100</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A2.** Performance Comparison over MHCflurry with MS (HR<sub>100</sub>; BLOSUM+Onehot+Deep)

Model	loss	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
ConvM	H <sub>v</sub>	5.72	6.89	3.15	5.66	7.11	12.78	10.81
	H <sub>l</sub>	4.89	5.54	2.47	3.50	6.27	7.34	8.35
	H <sub>i</sub>	4.75	5.88	0.93	3.62	6.24	7.21	7.29
	MS	-8.57	0.21	-8.00	-1.23	1.65	-2.45	-2.47
SpConvM	H <sub>v</sub>	9.86	<b>8.85</b>	<b>7.89</b>	8.59	<b>9.12</b>	<b>19.29</b>	<b>18.11</b>
	H <sub>l</sub>	8.18	7.64	6.46	6.64	8.44	15.59	14.08
	H <sub>i</sub>	7.35	8.5	5.02	6.41	7.96	13.21	14.04
	MS	7.10	8.05	2.49	4.95	6.30	15.17	12.5
MHCflurry	H <sub>v</sub>	<b>10.11</b>	8.54	7.14	<b>9.21</b>	6.09	16.00	14.65
	H <sub>l</sub>	8.78	6.60	5.55	5.95	5.50	13.19	11.30
	H <sub>i</sub>	6.50	6.20	5.08	5.47	4.69	12.19	8.12
	MS	0.00	0.00	0.00	0.00	0.00	0.00	0.00

The values in the table are percentage improvement compared with the baseline MHCflurry with MS. Models are trained using BLOSUM+Onehot+Deep encoding methods, and selected with respect to HR<sub>100</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A3.** Performance Comparison over MHCflurry with MS (AUC; BLOSUM+Onehot+Deep)

Model	loss	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
ConvM	H <sub>v</sub>	6.83	2.28	5.03	7.34	3.44	9.69	8.55
	H <sub>l</sub>	3.00	3.12	3.88	6.88	3.16	5.41	5.81
	H <sub>i</sub>	3.66	-0.63	3.01	5.83	2.81	4.43	3.58
	MS	-9.12	-8.45	-5.46	-2.31	-0.52	-5.62	-5.45
SpConvM	H <sub>v</sub>	<b>10.27</b>	<b>4.42</b>	<b>9.24</b>	<b>10.08</b>	<b>5.09</b>	<b>17.20</b>	<b>15.06</b>
	H <sub>l</sub>	7.18	2.84	8.12	8.31	4.42	12.36	10.63
	H <sub>i</sub>	7.54	2.98	7.69	7.25	4.44	11.57	10.04
	MS	4.31	2.49	5.28	4.78	2.86	11.68	9.47
MHCflurry	H <sub>v</sub>	8.08	1.61	7.45	8.35	4.30	11.02	11.05
	H <sub>l</sub>	7.64	3.96	6.31	7.61	3.56	8.63	7.63
	H <sub>i</sub>	7.39	1.64	5.81	6.74	3.41	9.32	7.68
	MS	0.00	0.00	0.00	0.00	0.00	0.00	0.00

The values in the table are percentage improvement compared with the baseline MHCflurry with MS. Models are trained using BLOSUM+Onehot+Deep encoding methods, and selected with respect to AUC and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A4.** Performance Comparison over MHCflurry with MS (ROC<sub>5</sub>; BLOSUM+Onehot+Deep)

Model	loss	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
ConvM	H <sub>v</sub>	8.38	6.71	3.23	6.86	5.14	5.91	4.63
	H <sub>l</sub>	3.33	5.16	0.96	5.52	4.22	5.22	2.56
	H <sub>i</sub>	8.36	7.95	3.35	5.83	5.01	5.40	3.70
	MS	-11.59	-5.75	-8.19	-2.14	0.23	-7.20	-8.84
SpConvM	H <sub>v</sub>	<b>12.78</b>	<b>15.81</b>	<b>8.31</b>	<b>11.11</b>	<b>7.29</b>	<b>15.03</b>	<b>10.52</b>
	H <sub>l</sub>	11.71	9.88	5.45	8.12	6.20	12.09	7.72
	H <sub>i</sub>	10.18	8.09	7.40	9.58	6.65	10.38	8.09
	MS	5.25	3.68	2.96	4.81	4.49	13.54	6.87
MHCflurry	H <sub>v</sub>	11.20	11.53	7.09	7.44	6.32	10.38	8.31
	H <sub>l</sub>	1.87	0.36	0.09	1.16	3.22	13.76	9.80
	H <sub>i</sub>	7.64	8.25	5.83	6.28	5.36	8.66	7.74
	MS	0.00	0.00	0.00	0.00	0.00	0.00	0.00

The values in the table are percentage improvement compared with the baseline MHCflurry with MS. Models are trained using BLOSUM+Onehot+Deep encoding methods, and selected with respect to ROC<sub>5</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A5.** Encoding Performance Comparison on SpConvM with H<sub>v</sub> (AR<sub>100</sub>)

encoding	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
BLOSUM	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Onehot	-6.04	-4.09	-6.00	-1.75	-1.13	-5.38	-5.34
Deep	-4.74	-0.98	-2.55	-0.09	-0.13	0.26	-1.17
BLOSUM +Onehot	-0.62	-0.06	-0.62	1.68	0.27	3.17	1.22
BLOSUM +Deep	<b>1.43</b>	<b>3.33</b>	<b>0.66</b>	2.62	<b>0.57</b>	4.88	2.73
Onehot +Deep	-4.10	-3.16	-2.64	-0.37	0.01	-1.83	-2.29
BLOSUM +Onehot +Deep	0.15	-0.15	-0.47	<b>3.53</b>	0.53	<b>6.29</b>	<b>3.53</b>

The values in the table are percentage improvement compared with the BLOSUM encoding method. Models are selected with respect to AR<sub>100</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A6.** Encoding Performance Comparison on SpConvM with  $H_v$  (HR<sub>100</sub>)

encoding	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
BLOSUM	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Onehot	-6.20	-3.31	-5.71	-3.85	-0.69	-7.10	-5.47
Deep	-3.81	-0.44	-2.64	-0.21	0.26	-0.36	0.09
BLOSUM +Onehot	-1.47	0.98	-0.70	-0.04	0.11	1.61	0.85
BLOSUM +Deep	-0.40	<b>2.45</b>	-0.30	0.57	0.28	3.39	1.97
Onehot +Deep	-6.45	-2.48	-4.34	-1.71	-0.55	-3.08	-3.13
BLOSUM +Onehot +Deep	<b>0.58</b>	0.07	<b>0.98</b>	<b>2.16</b>	<b>0.75</b>	<b>4.56</b>	<b>3.50</b>

The values in the table are percentage improvement compared with the BLOSUM encoding method. Models are selected with respect to HR<sub>100</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A7.** Encoding Performance Comparison on SpConvM with  $H_v$  (AUC)

encoding	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
BLOSUM	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Onehot	-6.75	-4.84	-4.37	-2.55	-0.91	-6.85	-5.34
Deep	-3.31	-0.59	-2.09	-0.71	-0.19	-1.06	-1.49
BLOSUM +Onehot	-0.45	<b>0.97</b>	0.22	0.52	0.09	1.41	1.02
BLOSUM +Deep	<b>0.95</b>	-0.84	1.09	<b>1.58</b>	0.55	3.25	2.00
Onehot +Deep	-2.90	-3.40	-2.68	-1.56	-0.39	-4.13	-3.26
BLOSUM +Onehot +Deep	0.77	-0.67	<b>1.43</b>	1.05	<b>0.62</b>	<b>5.90</b>	<b>3.95</b>

The values in the table are percentage improvement compared with the BLOSUM encoding method. Models are selected with respect to AUC and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.

**Table A8.** Encoding Performance Comparison on SpConvM with  $H_v$  (ROC<sub>5</sub>)

encoding	AR <sub>100</sub>	HR <sub>100</sub>	AR <sub>500</sub>	HR <sub>500</sub>	AUC	ROC <sub>5</sub>	ROC <sub>10</sub>
BLOSUM	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Onehot	-7.96	-7.56	-5.22	-2.32	-1.05	-5.61	-4.81
Deep	-2.29	-0.84	-2.76	-0.78	-0.54	-0.69	-1.28
BLOSUM +Onehot	1.24	-0.16	0.18	0.40	0.00	1.38	1.12
BLOSUM +Deep	<b>2.84</b>	0.15	1.41	<b>1.90</b>	0.60	3.20	3.14
Onehot +Deep	-4.76	-2.90	-3.31	-1.28	-0.61	-2.58	-2.30
BLOSUM +Onehot +Deep	2.57	<b>0.20</b>	<b>2.08</b>	1.46	<b>0.82</b>	<b>3.83</b>	<b>3.47</b>

The values in the table are percentage improvement compared with the BLOSUM encoding method. Models are selected with respect to ROC<sub>5</sub> and evaluated using the 7 evaluation metrics. The best improvement with respect to each metric is **bold**.