

## APPENDIX

In this appendix, we present the experimental results of our proposed model stability based stopping criterion for extended models including kernelized models and non-gradient-based models which are introduced in Section 7. We left these detailed experimental results for this appendix due to the page limitation (25 pages).

### A EXPERIMENTS FOR KERNELIZED MODELS

In this section, we validate our proposed algorithm on the kernelized models, including kernel logistic regression (kernel LR) and kernel SVMs with radial basis function (RBF) kernel and polynomial (PN) kernel (Section 7.1). Extensive experiments are implemented on both the UCI and ImageNet data sets with uncertainty sampling (US) strategy and expected model change (EMC) strategy, respectively.

#### A.1 Experiments with Uncertainty Sampling Strategy

We first present the experimental results coupled with US strategy. The following two subsections are detailed results for kernel LR and kernel SVMs, respectively.

*A.1.1 Experiments for Kernel LR.* Table 1-Table 2 present the results of kernel LR on UCI data sets with RBF kernel and PN kernel, respectively. Table 3-Table 4 present the results of kernel LR for ImageNet data sets. As shown in the tables, our proposed method MS is still observed to outperform the baselines most of the time. It achieves the highest accuracy with fewer annotations on most cases which demonstrates the effectiveness of our stopping criterion.

Table 1. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-LR with RBF kernel taking US as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Biodeg</b>	Size	52.0	446.0	843.0	$\Delta_{Size}$	<b>-101.0</b>	391.0	397.0	264.0	378.0	-88.0
	Acc	74.72%	87.45%	81.04%	$\Delta_{Acc}$	<b>1.7%</b>	6.32%	6.42%	2.08%	2.26%	2.08%
<b>Clave</b>	Size	430.0	1393.0	6880.0	$\Delta_{Size}$	<b>-86.0</b>	1135.0	5487.0	464.0	3354.0	344.0
	Acc	85.77%	98.42%	90.8%	$\Delta_{Acc}$	<b>0.2%</b>	0.28%	7.62%	0.29%	0.52%	0.47%
<b>Ionosphere</b>	Size	17.0	87.0	280.0	$\Delta_{Size}$	<b>-10.0</b>	-3.0	186.0	193.0	193.0	193.0
	Acc	75.21%	92.96%	84.79%	$\Delta_{Acc}$	<b>2.54%</b>	2.82%	8.45%	8.17%	8.17%	8.17%
<b>Spambase</b>	Size	230.0	1873.0	3680.0	$\Delta_{Size}$	<b>-407.0</b>	1308.0	1807.0	1807.0	1807.0	1807.0
	Acc	83.2%	94.26%	93.22%	$\Delta_{Acc}$	<b>0.28%</b>	0.81%	1.04%	1.04%	1.04%	1.04%
<b>WDBC</b>	Size	28.0	93.0	454.0	$\Delta_{Size}$	<b>-19.0</b>	5.0	283.0	101.0	138.0	23.0
	Acc	94.35%	98.26%	97.74%	$\Delta_{Acc}$	<b>0.43%</b>	0.87%	3.74%	0.7%	0.61%	3.3%
<b>D-vs-P</b>	Size	80.0	139.0	1286.0	$\Delta_{Size}$	<b>16.0</b>	22.0	1032.0	44.0	457.0	-59.0
	Acc	97.38%	99.64%	99.36%	$\Delta_{Acc}$	<b>0.14%</b>	0.17%	0.3%	0.22%	0.19%	2.27%
<b>E-vs-F</b>	Size	77.0	134.0	1234.0	$\Delta_{Size}$	<b>27.0</b>	<b>27.0</b>	1082.0	51.0	360.0	-42.0
	Acc	91.2%	99.61%	99.03%	$\Delta_{Acc}$	<b>0.06%</b>	0.13%	0.52%	0.19%	0.26%	2.14%
<b>M-vs-N</b>	Size	78.0	161.0	1259.0	$\Delta_{Size}$	<b>24.0</b>	30.0	1006.0	149.0	640.0	-83.0
	Acc	95.95%	99.4%	98.42%	$\Delta_{Acc}$	<b>0.25%</b>	0.47%	0.95%	0.51%	0.79%	3.45%
<b>U-vs-V</b>	Size	78.0	113.0	1260.0	$\Delta_{Size}$	<b>23.0</b>	32.0	1135.0	32.0	263.0	-31.0
	Acc	96.68%	99.92%	97.88%	$\Delta_{Acc}$	<b>0.02%</b>	<b>0.02%</b>	2.04%	<b>0.02%</b>	0.11%	1.58%

Table 2. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-LR with PN kernel taking US as the strategy. The thresholds of MS are 2.5, 2.0, 1.0, 2.5, 2.5, 2.0, 2.5, 2.0, 2.0 respectively.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Biodeg</b>	<i>Size</i>	52.0	272.0	843.0	$\Delta_{Size}$	<b>319.0</b>	484.0	571.0	561.0	571.0	429.0
	<i>Acc</i>	63.92%	89.39%	84.43%	$\Delta_{Acc}$	<b>2.83%</b>	4.25%	4.95%	4.72%	4.95%	3.54%
<b>Clave</b>	<i>Size</i>	430.0	1247.0	6880.0	$\Delta_{Size}$	839.0	<b>430.0</b>	4236.0	4386.0	5139.0	1226.0
	<i>Acc</i>	96.22%	97.91%	96.98%	$\Delta_{Acc}$	<b>0.19%</b>	0.23%	0.64%	0.9%	0.96%	0.33%
<b>Ionosphere</b>	<i>Size</i>	17.0	95.0	280.0	$\Delta_{Size}$	158.0	185.0	<b>128.0</b>	185.0	185.0	185.0
	<i>Acc</i>	73.24%	88.03%	84.51%	$\Delta_{Acc}$	<b>2.11%</b>	3.52%	2.82%	3.52%	3.52%	3.52%
<b>Spambase</b>	<i>Size</i>	230.0	1360.0	3680.0	$\Delta_{Size}$	1630.0	2129.0	<b>-789.0</b>	2320.0	2320.0	2320.0
	<i>Acc</i>	77.87%	88.04%	82.49%	$\Delta_{Acc}$	4.06%	5.94%	<b>2.7%</b>	5.55%	5.55%	5.55%
<b>WDBC</b>	<i>Size</i>	28.0	72.0	454.0	$\Delta_{Size}$	<b>18.0</b>	152.0	322.0	368.0	375.0	332.0
	<i>Acc</i>	86.96%	98.43%	94.26%	$\Delta_{Acc}$	<b>3.83%</b>	4.0%	4.09%	4.35%	4.26%	4.43%
<b>D-vs-P</b>	<i>Size</i>	80.0	160.0	1286.0	$\Delta_{Size}$	<b>133.0</b>	173.0	1018.0	1126.0	1126.0	1066.0
	<i>Acc</i>	95.0%	98.73%	96.18%	$\Delta_{Acc}$	<b>0.68%</b>	0.75%	2.36%	2.55%	2.55%	4.69%
<b>E-vs-F</b>	<i>Size</i>	77.0	272.0	1234.0	$\Delta_{Size}$	<b>-165.0</b>	465.0	962.0	540.0	803.0	45.0
	<i>Acc</i>	93.69%	97.9%	96.6%	$\Delta_{Acc}$	<b>0.49%</b>	0.81%	1.29%	1.46%	1.46%	0.65%
<b>M-vs-N</b>	<i>Size</i>	78.0	228.0	1259.0	$\Delta_{Size}$	<b>234.0</b>	360.0	1031.0	1022.0	1029.0	848.0
	<i>Acc</i>	93.99%	97.11%	95.89%	$\Delta_{Acc}$	<b>0.59%</b>	<b>0.59%</b>	1.23%	1.74%	1.23%	1.03%
<b>U-vs-V</b>	<i>Size</i>	78.0	259.0	1260.0	$\Delta_{Size}$	<b>128.0</b>	736.0	1001.0	896.0	996.0	-181.0
	<i>Acc</i>	96.85%	99.68%	97.9%	$\Delta_{Acc}$	<b>0.32%</b>	1.47%	1.79%	2.31%	1.89%	2.84%

Table 3. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-LR with RBF kernel taking US as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Basenji</b>	Size	184.0	2108.0	2951.0	$\Delta_{Size}$	<b>-222.0</b>	-481.0	518.0	592.0	407.0	703.0
	Acc	55.48%	65.49%	64.41%	$\Delta_{Acc}$	<b>0.68%</b>	2.71%	2.84%	1.22%	1.35%	1.62%
<b>Chihuahua</b>	Size	175.0	2275.0	2800.0	$\Delta_{Size}$	<b>-945.0</b>	-525.0	105.0	-140.0	455.0	70.0
	Acc	47.29%	69.14%	67.43%	$\Delta_{Acc}$	<b>0.43%</b>	2.0%	1.0%	1.14%	2.0%	0.57%
<b>English setter</b>	Size	242.0	1712.0	3881.0	$\Delta_{Size}$	<b>-490.0</b>	-686.0	196.0	931.0	1617.0	1813.0
	Acc	59.94%	64.68%	63.44%	$\Delta_{Acc}$	<b>1.44%</b>	2.27%	2.16%	2.37%	2.57%	1.96%
<b>German police dog</b>	Size	174.0	1994.0	2785.0	$\Delta_{Size}$	<b>-595.0</b>	-385.0	-280.0	420.0	490.0	70.0
	Acc	49.64%	64.71%	63.27%	$\Delta_{Acc}$	<b>0.14%</b>	0.72%	1.29%	1.15%	1.72%	3.59%
<b>Standard poodle</b>	Size	195.0	3120.0	3129.0	$\Delta_{Size}$	<b>-1170.0</b>	-1326.0	-351.0	-117.0	-78.0	-1287.0
	Acc	45.08%	75.99%	74.2%	$\Delta_{Acc}$	<b>1.4%</b>	2.04%	2.17%	2.17%	2.55%	2.55%
<b>Vizsla</b>	Size	233.0	3288.0	3734.0	$\Delta_{Size}$	282.0	47.0	-1034.0	<b>94.0</b>	188.0	-846.0
	Acc	60.28%	70.24%	69.7%	$\Delta_{Acc}$	<b>1.39%</b>	3.1%	2.25%	3.32%	2.89%	1.61%
<b>Yorkshire terrier</b>	Size	304.0	4574.0	4874.0	$\Delta_{Size}$	<b>-305.0</b>	-1647.0	300.0	-61.0	183.0	-1098.0
	Acc	68.28%	75.49%	74.84%	$\Delta_{Acc}$	<b>0.16%</b>	0.9%	0.66%	1.39%	2.13%	0.74%
<b>Wild dog</b>	Size	242.0	2594.0	3875.0	$\Delta_{Size}$	960.0	<b>-144.0</b>	-720.0	384.0	480.0	864.0
	Acc	65.84%	72.86%	72.03%	$\Delta_{Acc}$	<b>0.31%</b>	0.62%	1.65%	3.61%	0.52%	1.75%
<b>Wolf</b>	Size	561.0	5713.0	8986.0	$\Delta_{Size}$	<b>896.0</b>	-1904.0	-3839.0	1568.0	1680.0	-1232.0
	Acc	72.33%	76.87%	74.6%	$\Delta_{Acc}$	3.29%	1.47%	5.66%	<b>0.53%</b>	2.58%	1.65%
<b>Fox</b>	Size	478.0	7582.0	7654.0	$\Delta_{Size}$	-576.0	-1536.0	72.0	<b>-192.0</b>	<b>-192.0</b>	-1248.0
	Acc	72.52%	77.69%	76.96%	$\Delta_{Acc}$	0.37%	0.57%	0.73%	<b>0.05%</b>	<b>0.05%</b>	0.21%
<b>Rabbit</b>	Size	349.0	3709.0	5593.0	$\Delta_{Size}$	980.0	-1190.0	1050.0	1750.0	1884.0	<b>-490.0</b>
	Acc	69.12%	76.34%	74.2%	$\Delta_{Acc}$	<b>0.64%</b>	0.86%	2.29%	3.57%	2.14%	1.14%
<b>Cat</b>	Size	732.0	7788.0	11722.0	$\Delta_{Size}$	<b>3731.0</b>	3934.0	3934.0	3934.0	3934.0	3934.0
	Acc	64.68%	69.78%	65.98%	$\Delta_{Acc}$	<b>2.08%</b>	3.8%	3.8%	3.8%	3.8%	3.8%
<b>Panda</b>	Size	351.0	2871.0	5628.0	$\Delta_{Size}$	<b>1540.0</b>	-2030.0	1820.0	1680.0	2240.0	2170.0
	Acc	55.61%	84.52%	81.04%	$\Delta_{Acc}$	<b>0.14%</b>	2.13%	2.98%	0.92%	2.27%	2.77%
<b>Elephant</b>	Size	392.0	3710.0	6282.0	$\Delta_{Size}$	<b>316.0</b>	-2844.0	2572.0	869.0	1738.0	-2607.0
	Acc	80.47%	85.75%	83.21%	$\Delta_{Acc}$	<b>0.25%</b>	2.42%	2.54%	0.57%	1.91%	3.69%

Table 4. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-LR with PN kernel taking US as the strategy. The thresholds of MS are all 2.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Basenji</b>	Size	184.0	1689.0	2951.0	$\Delta_{Size}$	<b>259.0</b>	1262.0	-604.0	1262.0	1262.0	1262.0
	Acc	51.96%	62.02%	59.27%	$\Delta_{Acc}$	<b>1.58%</b>	2.75%	6.0%	2.75%	2.75%	2.75%
<b>Chihuahua</b>	Size	175.0	1692.0	2800.0	$\Delta_{Size}$	<b>560.0</b>	1108.0	-1027.0	1108.0	1108.0	1108.0
	Acc	55.29%	63.9%	63.1%	$\Delta_{Acc}$	<b>0.48%</b>	0.81%	7.52%	0.81%	0.81%	0.81%
<b>English setter</b>	Size	242.0	2864.0	3881.0	$\Delta_{Size}$	<b>573.0</b>	<b>573.0</b>	-1923.0	1018.0	1018.0	1018.0
	Acc	53.58%	61.89%	60.04%	$\Delta_{Acc}$	<b>1.57%</b>	4.76%	5.79%	1.85%	1.85%	1.85%
<b>German police dog</b>	Size	174.0	1376.0	2785.0	$\Delta_{Size}$	828.0	1409.0	<b>128.0</b>	1409.0	1409.0	1409.0
	Acc	55.76%	61.88%	59.4%	$\Delta_{Acc}$	<b>1.24%</b>	2.49%	5.88%	2.49%	2.49%	2.49%
<b>Standard poodle</b>	Size	195.0	2291.0	3129.0	$\Delta_{Size}$	<b>419.0</b>	838.0	-1446.0	838.0	838.0	838.0
	Acc	62.41%	70.29%	69.31%	$\Delta_{Acc}$	<b>0.68%</b>	0.98%	0.81%	0.98%	0.98%	0.98%
<b>Vizsla</b>	Size	233.0	2442.0	3734.0	$\Delta_{Size}$	<b>928.0</b>	1292.0	-1927.0	1292.0	1292.0	1292.0
	Acc	55.41%	64.88%	62.26%	$\Delta_{Acc}$	<b>1.5%</b>	2.62%	5.84%	2.62%	2.62%	2.62%
<b>Yorkshire terrier</b>	Size	304.0	3212.0	4874.0	$\Delta_{Size}$	<b>-1891.0</b>	1662.0	324.0	1662.0	1662.0	1662.0
	Acc	65.22%	70.57%	68.58%	$\Delta_{Acc}$	<b>1.07%</b>	1.99%	7.62%	1.99%	1.99%	1.99%
<b>Wild dog</b>	Size	242.0	2450.0	3875.0	$\Delta_{Size}$	<b>528.0</b>	1425.0	-1344.0	1425.0	1425.0	1425.0
	Acc	57.89%	71.1%	66.87%	$\Delta_{Acc}$	<b>1.14%</b>	4.23%	1.34%	4.23%	4.23%	4.23%
<b>Wolf</b>	Size	561.0	5601.0	8986.0	$\Delta_{Size}$	<b>-1904.0</b>	3385.0	3385.0	3385.0	3385.0	3385.0
	Acc	64.41%	70.6%	68.37%	$\Delta_{Acc}$	<b>1.56%</b>	2.22%	2.22%	2.22%	2.22%	2.22%
<b>Fox</b>	Size	478.0	1918.0	7654.0	$\Delta_{Size}$	3072.0	5736.0	<b>2592.0</b>	5736.0	5736.0	5736.0
	Acc	65.57%	70.74%	69.38%	$\Delta_{Acc}$	<b>0.31%</b>	1.36%	0.73%	1.36%	1.36%	1.36%
<b>Rabbit</b>	Size	349.0	4164.0	5593.0	$\Delta_{Size}$	<b>-1348.0</b>	1429.0	330.0	1429.0	1429.0	1429.0
	Acc	55.5%	69.1%	66.14%	$\Delta_{Acc}$	<b>2.2%</b>	2.97%	6.67%	2.97%	2.97%	2.97%
<b>Cat</b>	Size	732.0	10471.0	11722.0	$\Delta_{Size}$	-9200.0	<b>1251.0</b>	-5672.0	<b>1251.0</b>	<b>1251.0</b>	<b>1251.0</b>
	Acc	49.62%	65.78%	65.39%	$\Delta_{Acc}$	4.25%	<b>0.39%</b>	0.76%	<b>0.39%</b>	<b>0.39%</b>	<b>0.39%</b>
<b>Panda</b>	Size	351.0	3396.0	5628.0	$\Delta_{Size}$	<b>679.0</b>	2232.0	1869.0	2232.0	2232.0	2232.0
	Acc	75.96%	77.77%	76.17%	$\Delta_{Acc}$	<b>1.03%</b>	1.6%	1.1%	1.6%	1.6%	1.6%
<b>Elephant</b>	Size	392.0	3434.0	6282.0	$\Delta_{Size}$	970.0	2849.0	<b>536.0</b>	2849.0	2849.0	2849.0
	Acc	69.94%	81.3%	77.89%	$\Delta_{Acc}$	<b>2.42%</b>	3.4%	3.56%	3.4%	3.4%	3.4%

*A.1.2 Experiments for Kernel SVM.* Table 5-Table 6 present the results of kernel SVM on UCI data sets with RBF kernel and PN kernel, respectively. Table 7-Table 8 are the results of kernel SVM on ImageNet data sets. From these tables, similar results can be observed and MS performs best in these seven stopping criteria.

Table 5. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-SVM with RBF kernel taking US as the strategy. The thresholds of MS are 2.0, 2.0, 2.0, 2.0, 2.5, 2.0, 2.0, 2.0, 2.5 respectively.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME	QUIRE
Biodeg	Size	52.0	250.0	843.0	$\Delta_{Size}$	107.0	152.0	240.0	367.0	501.0	84.0	<b>-29.0</b>
	Acc	77.6%	88.36%	86.87%	$\Delta_{Acc}$	<b>1.19%</b>	1.43%	1.75%	1.58%	1.6%	1.38%	2.68%
Clave	Size	430.0	2620.0	6880.0	$\Delta_{Size}$	-1250.0	-1221.0	350.0	-1456.0	-1193.0	-2190.0	<b>-957.0</b>
	Acc	96.16%	98.33%	98.21%	$\Delta_{Acc}$	0.13%	0.14%	0.16%	0.29%	0.14%	2.17%	<b>0.11%</b>
Ionosphere	Size	17.0	64.0	280.0	$\Delta_{Size}$	161.0	40.0	99.0	61.0	136.0	27.0	<b>6.0</b>
	Acc	78.73%	95.92%	94.65%	$\Delta_{Acc}$	<b>1.06%</b>	1.76%	1.34%	1.27%	1.34%	1.97%	8.24%
Spambase	Size	230.0	1423.0	3680.0	$\Delta_{Size}$	<b>-199.0</b>	-330.0	1708.0	-250.0	880.0	-391.0	343.0
	Acc	87.36%	93.53%	93.16%	$\Delta_{Acc}$	0.61%	0.43%	<b>0.42%</b>	0.47%	0.25%	0.53%	0.45%
WDBC	Size	28.0	64.0	454.0	$\Delta_{Size}$	366.0	40.0	318.0	28.0	80.0	<b>23.0</b>	101.0
	Acc	94.49%	98.72%	98.2%	$\Delta_{Acc}$	<b>0.58%</b>	0.81%	0.75%	0.99%	1.04%	0.87%	0.93%
D-vs-P	Size	80.0	148.0	1286.0	$\Delta_{Size}$	26.0	<b>18.0</b>	629.0	32.0	98.0	-68.0	152.0
	Acc	96.54%	99.53%	99.35%	$\Delta_{Acc}$	<b>0.11%</b>	0.28%	0.17%	0.19%	0.17%	3.0%	0.26%
E-vs-F	Size	77.0	184.0	1234.0	$\Delta_{Size}$	32.0	24.0	540.0	<b>19.0</b>	86.0	512.0	94.0
	Acc	94.94%	99.34%	99.06%	$\Delta_{Acc}$	<b>0.16%</b>	0.26%	0.29%	0.29%	0.32%	0.84%	0.33%
M-vs-N	Size	78.0	171.0	1259.0	$\Delta_{Size}$	380.0	<b>48.0</b>	707.0	54.0	163.0	-93.0	109.0
	Acc	96.65%	98.99%	98.8%	$\Delta_{Acc}$	<b>0.19%</b>	0.32%	0.25%	0.25%	0.25%	2.34%	0.28%
U-vs-V	Size	78.0	140.0	1260.0	$\Delta_{Size}$	285.0	22.0	601.0	<b>10.0</b>	89.0	-62.0	68.0
	Acc	97.05%	99.72%	99.57%	$\Delta_{Acc}$	0.57%	0.16%	0.14%	<b>0.09%</b>	0.14%	2.67%	0.11%

Table 6. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-SVM with PN kernel taking US as the strategy. The thresholds of MS are 2.0, 2.0, 2.0, 1.5, 2.0, 2.0, 2.0, 2.0, 2.0 respectively.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME	QUIRE
Biodeg	Size	52.0	339.0	843.0	$\Delta_{Size}$	-108.0	-33.0	43.0	66.0	271.0	<b>-50.0</b>	-103.0
	Acc	77.78%	89.15%	87.69%	$\Delta_{Acc}$	1.42%	1.51%	2.26%	1.65%	1.65%	<b>1.27%</b>	1.46%
Clave	Size	430.0	1238.0	6880.0	$\Delta_{Size}$	<b>-17.0</b>	-172.0	1926.0	-52.0	559.0	-808.0	163.0
	Acc	94.19%	97.85%	96.33%	$\Delta_{Acc}$	<b>0.23%</b>	0.31%	0.9%	0.29%	0.52%	3.66%	0.37%
Ionosphere	Size	17.0	102.0	280.0	$\Delta_{Size}$	<b>-10.0</b>	5.0	45.0	107.0	151.0	64.0	-17.0
	Acc	72.58%	90.23%	84.23%	$\Delta_{Acc}$	<b>3.47%</b>	4.41%	4.88%	5.63%	5.45%	4.88%	4.41%
Spambase	Size	230.0	2134.0	3680.0	$\Delta_{Size}$	-1104.0	-1063.0	934.0	-920.0	<b>23.0</b>	-1251.0	-1086.0
	Acc	87.08%	93.45%	93.11%	$\Delta_{Acc}$	0.8%	0.65%	<b>0.47%</b>	0.54%	0.66%	0.66%	0.67%
WDBC	Size	28.0	60.0	454.0	$\Delta_{Size}$	<b>0.0</b>	23.0	302.0	45.0	122.0	50.0	48.0
	Acc	93.45%	98.09%	96.81%	$\Delta_{Acc}$	<b>0.7%</b>	1.16%	1.22%	1.28%	0.99%	1.16%	1.28%
D-vs-P	Size	80.0	223.0	1286.0	$\Delta_{Size}$	-47.0	-58.0	510.0	-54.0	70.0	-93.0	<b>1.0</b>
	Acc	96.27%	99.42%	99.25%	$\Delta_{Acc}$	0.35%	0.35%	<b>0.08%</b>	0.35%	0.27%	1.51%	0.33%
E-vs-F	Size	77.0	160.0	1234.0	$\Delta_{Size}$	18.0	<b>8.0</b>	333.0	33.0	102.0	1075.0	57.0
	Acc	96.6%	99.48%	99.19%	$\Delta_{Acc}$	<b>0.19%</b>	0.26%	0.39%	0.23%	<b>0.19%</b>	0.29%	0.23%
M-vs-N	Size	78.0	184.0	1259.0	$\Delta_{Size}$	24.0	<b>-5.0</b>	498.0	83.0	131.0	1075.0	70.0
	Acc	95.32%	99.68%	99.43%	$\Delta_{Acc}$	<b>0.16%</b>	0.7%	0.19%	0.25%	0.19%	0.25%	0.54%
U-vs-V	Size	78.0	136.0	1260.0	$\Delta_{Size}$	28.0	27.0	492.0	<b>25.0</b>	82.0	-58.0	113.0
	Acc	97.01%	99.75%	99.62%	$\Delta_{Acc}$	<b>0.13%</b>	0.17%	0.17%	0.19%	0.17%	2.73%	0.15%

Table 7. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-SVM with RBF kernel taking US as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME	QUIRE
Basenji	Size	184.0	1897.0	2951.0	$\Delta_{Size}$	902.0	844.0	-453.0	1054.0	1054.0	1054.0	-370.0
	Acc	55.87%	64.19%	63.33%	$\Delta_{Acc}$	0.72%	0.89%	3.27%	0.87%	0.87%	0.87%	1.79%
Chihuahua	Size	175.0	1829.0	2800.0	$\Delta_{Size}$	166.0	464.0	-936.0	971.0	971.0	963.0	-359.0
	Acc	58.61%	65.21%	63.96%	$\Delta_{Acc}$	0.93%	1.29%	2.25%	1.25%	1.25%	1.29%	2.47%
English setter	Size	242.0	2496.0	3881.0	$\Delta_{Size}$	743.0	1209.0	-1372.0	1385.0	1385.0	1383.0	-1421.0
	Acc	56.42%	65.5%	64.04%	$\Delta_{Acc}$	1.07%	1.36%	4.43%	1.45%	1.45%	1.44%	3.98%
German police dog	Size	174.0	1959.0	2785.0	$\Delta_{Size}$	410.0	689.0	-899.0	826.0	826.0	826.0	-1155.0
	Acc	56.89%	63.79%	63.07%	$\Delta_{Acc}$	0.63%	0.95%	3.32%	0.72%	0.72%	0.72%	3.6%
Standard poodle	Size	195.0	2188.0	3129.0	$\Delta_{Size}$	0.0	-47.0	-120.0	917.0	933.0	-94.0	-733.0
	Acc	65.7%	74.64%	73.98%	$\Delta_{Acc}$	0.6%	0.75%	1.34%	0.66%	0.65%	0.92%	2.25%
Vizsla	Size	233.0	3175.0	3734.0	$\Delta_{Size}$	-165.0	-66.0	-1222.0	559.0	559.0	348.0	-780.0
	Acc	59.83%	68.63%	68.07%	$\Delta_{Acc}$	0.41%	0.66%	3.1%	0.56%	0.56%	0.64%	1.09%
Yorkshire terrier	Size	304.0	3385.0	4874.0	$\Delta_{Size}$	-299.0	-403.0	-868.0	1366.0	1450.0	-220.0	-610.0
	Acc	68.61%	74.81%	74.52%	$\Delta_{Acc}$	0.82%	0.83%	1.89%	0.26%	0.3%	0.66%	0.66%
Wild dog	Size	242.0	2623.0	3875.0	$\Delta_{Size}$	67.0	-422.0	202.0	1252.0	1252.0	403.0	-797.0
	Acc	66.5%	73.79%	73.23%	$\Delta_{Acc}$	0.52%	0.76%	0.76%	0.56%	0.56%	0.33%	0.99%
Wolf	Size	561.0	5567.0	8986.0	$\Delta_{Size}$	302.0	-829.0	2677.0	3270.0	3394.0	-314.0	784.0
	Acc	72.0%	77.39%	77.1%	$\Delta_{Acc}$	0.52%	0.54%	0.51%	0.28%	0.28%	0.39%	0.37%
Fox	Size	478.0	6027.0	7654.0	$\Delta_{Size}$	-1622.0	-1853.0	146.0	1555.0	1620.0	-1229.0	-749.0
	Acc	71.85%	77.62%	77.4%	$\Delta_{Acc}$	0.71%	0.76%	0.44%	0.22%	0.21%	0.62%	0.34%
Rabbit	Size	349.0	4138.0	5593.0	$\Delta_{Size}$	-56.0	-830.0	-427.0	1455.0	1455.0	-485.0	-737.0
	Acc	69.26%	77.15%	76.82%	$\Delta_{Acc}$	0.28%	0.52%	1.42%	0.33%	0.33%	0.42%	0.73%
Cat	Size	732.0	7964.0	11722.0	$\Delta_{Size}$	2587.0	-735.0	626.0	3758.0	3758.0	2705.0	1264.0
	Acc	64.69%	69.09%	68.77%	$\Delta_{Acc}$	0.14%	0.46%	0.79%	0.33%	0.33%	0.24%	0.29%
Panda	Size	351.0	2570.0	5628.0	$\Delta_{Size}$	49.0	119.0	2255.0	2296.0	2709.0	-105.0	490.0
	Acc	77.7%	83.03%	82.57%	$\Delta_{Acc}$	0.26%	0.36%	0.65%	0.48%	0.43%	0.43%	0.42%
Elephant	Size	392.0	3355.0	6282.0	$\Delta_{Size}$	-474.0	-474.0	1128.0	1264.0	2094.0	-869.0	-382.0
	Acc	78.94%	84.32%	84.13%	$\Delta_{Acc}$	0.57%	0.54%	0.48%	0.1%	0.19%	0.89%	0.2%

Table 8. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-SVM with PN kernel taking US as the strategy. The thresholds of MS are 1.0, 1.0, 1.0, 1.0, 1.0, 1.0, 1.0, 0.5, 1.0, 1.0, 0.5, 1.0, 0.5, 0.5 respectively.

Dataset	Initial	HPP	All	MS	SP	GRAD	OU	MC	ME	QUIRE
<b>Basenji</b>	Size 184.0	2182.0	2951.0	$\Delta_{Size}$ <b>-229.0</b>	769.0	-1354.0	769.0	769.0	725.0	-1177.0
	Acc 55.51%	65.12%	63.9%	$\Delta_{Acc}$ <b>1.0%</b>	1.22%	3.27%	1.22%	1.22%	1.65%	2.98%
<b>Chihuahua</b>	Size 175.0	1848.0	2800.0	$\Delta_{Size}$ <b>-7.0</b>	875.0	-490.0	952.0	952.0	896.0	-840.0
	Acc 55.83%	66.11%	64.43%	$\Delta_{Acc}$ <b>0.83%</b>	1.6%	1.74%	1.69%	1.69%	1.74%	4.71%
<b>English setter</b>	Size 242.0	2625.0	3881.0	$\Delta_{Size}$ <b>121.0</b>	1256.0	-443.0	1256.0	1256.0	1231.0	-1187.0
	Acc 52.5%	64.32%	62.84%	$\Delta_{Acc}$ <b>0.84%</b>	1.47%	2.08%	1.47%	1.47%	1.73%	2.29%
<b>German police dog</b>	Size 174.0	2141.0	2785.0	$\Delta_{Size}$ <b>7.0</b>	220.0	-1130.0	644.0	644.0	636.0	-994.0
	Acc 52.8%	64.36%	62.61%	$\Delta_{Acc}$ <b>1.69%</b>	3.53%	6.05%	1.75%	1.75%	2.09%	3.3%
<b>Standard poodle</b>	Size 195.0	1762.0	3129.0	$\Delta_{Size}$ <b>-52.0</b>	169.0	187.0	1320.0	1362.0	137.0	-806.0
	Acc 62.01%	75.31%	74.24%	$\Delta_{Acc}$ <b>0.4%</b>	1.06%	2.23%	1.0%	1.21%	0.7%	3.19%
<b>Vizsla</b>	Size 233.0	3344.0	3734.0	$\Delta_{Size}$ -893.0	390.0	-715.0	380.0	390.0	<b>-75.0</b>	-1626.0
	Acc 60.45%	70.75%	69.81%	$\Delta_{Acc}$ 1.71%	0.94%	1.8%	1.13%	0.94%	<b>0.73%</b>	3.34%
<b>Yorkshire terrier</b>	Size 304.0	3064.0	4874.0	$\Delta_{Size}$ <b>-442.0</b>	-336.0	104.0	1571.0	1781.0	31.0	-671.0
	Acc 69.24%	75.86%	74.67%	$\Delta_{Acc}$ <b>0.55%</b>	0.74%	1.48%	0.72%	1.05%	1.17%	1.09%
<b>Wild dog</b>	Size 242.0	3346.0	3875.0	$\Delta_{Size}$ <b>-1216.0</b>	-1216.0	-1269.0	464.0	529.0	-720.0	-1232.0
	Acc 64.71%	74.48%	73.82%	$\Delta_{Acc}$ <b>0.48%</b>	0.76%	2.17%	0.79%	0.65%	0.86%	1.2%
<b>Wolf</b>	Size 561.0	6833.0	8986.0	$\Delta_{Size}$ <b>-2296.0</b>	-2128.0	2153.0	1792.0	2128.0	-1624.0	-1624.0
	Acc 73.0%	77.36%	76.69%	$\Delta_{Acc}$ <b>0.4%</b>	0.53%	0.67%	0.67%	0.73%	0.44%	0.67%
<b>Fox</b>	Size 478.0	6595.0	7654.0	$\Delta_{Size}$ -2811.0	-3099.0	1059.0	<b>880.0</b>	1035.0	-2448.0	-1915.0
	Acc 71.26%	77.77%	77.31%	$\Delta_{Acc}$ 1.02%	1.04%	<b>0.45%</b>	0.53%	0.57%	0.88%	0.75%
<b>Rabbit</b>	Size 349.0	3196.0	5593.0	$\Delta_{Size}$ 117.0	<b>23.0</b>	2397.0	2287.0	2397.0	373.0	607.0
	Acc 69.72%	77.25%	76.41%	$\Delta_{Acc}$ <b>0.21%</b>	0.91%	0.83%	1.0%	0.83%	0.41%	0.48%
<b>Cat</b>	Size 732.0	9589.0	11722.0	$\Delta_{Size}$ -2242.0	-2438.0	2133.0	2133.0	2133.0	<b>257.0</b>	-723.0
	Acc 64.44%	69.68%	69.41%	$\Delta_{Acc}$ 0.47%	0.34%	<b>0.27%</b>	<b>0.27%</b>	<b>0.27%</b>	0.34%	0.4%
<b>Panda</b>	Size 351.0	3137.0	5628.0	$\Delta_{Size}$ <b>-756.0</b>	-686.0	1199.0	1531.0	2114.0	-1246.0	-476.0
	Acc 76.09%	83.43%	82.67%	$\Delta_{Acc}$ <b>0.31%</b>	0.4%	0.47%	0.64%	0.66%	0.88%	0.4%
<b>Elephant</b>	Size 392.0	2525.0	6282.0	$\Delta_{Size}$ 119.0	<b>40.0</b>	3757.0	2173.0	3002.0	119.0	-79.0
	Acc 79.07%	85.34%	84.8%	$\Delta_{Acc}$ <b>0.35%</b>	0.38%	0.54%	0.64%	0.64%	0.41%	0.57%

## A.2 Experiments with Expected Model Change Strategy

In this section, we present the experimental results coupled with EMC sampling strategy. The following two subsections detail the results for kernel LR and kernel SVMs, respectively.

*A.2.1 Experiments for Kernel LR.* Table 9-Table 10 present the results of kernel LR on UCI data sets with RBF kernel and PN kernel, respectively. Table 11-Table 12 are the results on ImageNet data sets. As shown in these tables, MS still performs the best on most cases which is consistent with previous experiments.

Table 9. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-LR with RBF kernel taking EMC as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Biodeg</b>	Size	52.0	466.0	843.0	$\Delta_{Size}$	<b>-64.0</b>	377.0	336.0	377.0	377.0	276.0
	Acc	82.92%	90.94%	88.77%	$\Delta_{Acc}$	<b>1.7%</b>	2.17%	2.26%	2.17%	2.17%	1.98%
<b>Clave</b>	Size	430.0	1565.0	6880.0	$\Delta_{Size}$	447.0	894.0	5315.0	5315.0	5315.0	<b>155.0</b>
	Acc	87.03%	98.4%	96.88%	$\Delta_{Acc}$	<b>0.15%</b>	0.22%	1.51%	1.51%	1.51%	2.47%
<b>Ionosphere</b>	Size	17.0	123.0	280.0	$\Delta_{Size}$	<b>-29.0</b>	-6.0	157.0	157.0	157.0	157.0
	Acc	82.25%	93.52%	87.61%	$\Delta_{Acc}$	<b>3.1%</b>	<b>3.1%</b>	5.92%	5.92%	5.92%	5.92%
<b>Spambase</b>	Size	230.0	1324.0	3680.0	$\Delta_{Size}$	<b>261.0</b>	2075.0	2259.0	2356.0	2356.0	2356.0
	Acc	87.92%	94.14%	92.81%	$\Delta_{Acc}$	<b>0.37%</b>	1.31%	4.67%	1.33%	1.33%	1.33%
<b>WDBC</b>	Size	28.0	57.0	454.0	$\Delta_{Size}$	<b>20.0</b>	32.0	323.0	384.0	394.0	-29.0
	Acc	94.26%	98.78%	97.04%	$\Delta_{Acc}$	<b>1.04%</b>	1.57%	2.43%	1.57%	1.74%	4.52%
<b>D-vs-P</b>	Size	80.0	176.0	1286.0	$\Delta_{Size}$	<b>0.0</b>	-22.0	996.0	1108.0	1109.0	-3.0
	Acc	83.11%	99.57%	97.02%	$\Delta_{Acc}$	<b>0.12%</b>	0.25%	0.43%	0.5%	0.5%	0.25%
<b>E-vs-F</b>	Size	77.0	299.0	1234.0	$\Delta_{Size}$	<b>-123.0</b>	-120.0	860.0	925.0	934.0	-222.0
	Acc	95.34%	99.61%	99.16%	$\Delta_{Acc}$	<b>0.26%</b>	<b>0.26%</b>	0.45%	4.27%	4.27%	4.27%
<b>M-vs-N</b>	Size	78.0	163.0	1259.0	$\Delta_{Size}$	37.0	<b>19.0</b>	965.0	1091.0	1096.0	-85.0
	Acc	95.46%	99.31%	97.26%	$\Delta_{Acc}$	<b>0.05%</b>	<b>0.05%</b>	2.0%	1.0%	2.06%	3.85%
<b>U-vs-V</b>	Size	78.0	129.0	1260.0	$\Delta_{Size}$	<b>22.0</b>	26.0	787.0	1131.0	1131.0	-51.0
	Acc	97.92%	99.94%	99.68%	$\Delta_{Acc}$	<b>0.0%</b>	<b>0.0%</b>	0.25%	0.25%	0.25%	2.02%



Table 10. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-LR with PN kernel taking EMC as the strategy. The thresholds of MS are 1.0, 1.5, 1.0, 1.5, 1.0, 2.0, 2.0, 2.0, 2.0 respectively.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Biodeg</b>	<i>Size</i>	52.0	195.0	843.0	$\Delta_{Size}$	<b>275.0</b>	468.0	357.0	648.0	648.0	648.0
	<i>Acc</i>	78.77%	86.56%	84.91%	$\Delta_{Acc}$	<b>0.47%</b>	1.18%	1.18%	1.65%	1.65%	1.65%
<b>Clave</b>	<i>Size</i>	430.0	1388.0	6880.0	$\Delta_{Size}$	848.0	<b>160.0</b>	5492.0	4263.0	4902.0	885.0
	<i>Acc</i>	95.59%	97.98%	97.34%	$\Delta_{Acc}$	<b>0.22%</b>	0.29%	0.64%	0.66%	0.7%	0.75%
<b>Ionosphere</b>	<i>Size</i>	17.0	103.0	280.0	$\Delta_{Size}$	161.0	<b>116.0</b>	166.0	177.0	177.0	146.0
	<i>Acc</i>	69.53%	86.81%	83.87%	$\Delta_{Acc}$	<b>2.3%</b>	2.56%	4.1%	4.61%	2.94%	6.02%
<b>Spambase</b>	<i>Size</i>	230.0	543.0	3680.0	$\Delta_{Size}$	<b>-244.0</b>	856.0	2111.0	3137.0	3137.0	3137.0
	<i>Acc</i>	78.33%	85.73%	81.89%	$\Delta_{Acc}$	<b>2.58%</b>	5.07%	5.66%	3.84%	3.84%	3.84%
<b>WDBC</b>	<i>Size</i>	28.0	272.0	454.0	$\Delta_{Size}$	<b>-40.0</b>	-74.0	182.0	182.0	182.0	140.0
	<i>Acc</i>	85.22%	92.46%	91.3%	$\Delta_{Acc}$	<b>0.29%</b>	2.61%	1.16%	1.16%	1.16%	1.16%
<b>D-vs-P</b>	<i>Size</i>	80.0	244.0	1286.0	$\Delta_{Size}$	<b>-9.0</b>	101.0	623.0	1042.0	1042.0	997.0
	<i>Acc</i>	95.48%	98.59%	96.45%	$\Delta_{Acc}$	<b>0.62%</b>	0.69%	1.38%	2.14%	2.14%	2.31%
<b>E-vs-F</b>	<i>Size</i>	77.0	215.0	1234.0	$\Delta_{Size}$	<b>75.0</b>	306.0	1019.0	783.0	945.0	135.0
	<i>Acc</i>	95.02%	97.93%	95.79%	$\Delta_{Acc}$	<b>0.58%</b>	0.91%	2.14%	1.75%	1.88%	1.04%
<b>M-vs-N</b>	<i>Size</i>	78.0	242.0	1259.0	$\Delta_{Size}$	<b>30.0</b>	148.0	872.0	1017.0	1017.0	1017.0
	<i>Acc</i>	92.44%	97.27%	95.57%	$\Delta_{Acc}$	<b>0.59%</b>	0.79%	1.62%	1.7%	1.7%	1.7%
<b>U-vs-V</b>	<i>Size</i>	78.0	144.0	1260.0	$\Delta_{Size}$	<b>54.0</b>	66.0	1030.0	1116.0	1116.0	1083.0
	<i>Acc</i>	96.25%	98.83%	96.5%	$\Delta_{Acc}$	<b>0.5%</b>	0.63%	2.24%	2.33%	2.33%	2.33%

Table 11. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-LR with RBF kernel taking EMC as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Basenji</b>	Size	184.0	2774.0	2951.0	$\Delta_{Size}$	148.0	-1443.0	-370.0	<b>111.0</b>	177.0	-1036.0
	Acc	53.45%	64.55%	61.43%	$\Delta_{Acc}$	<b>0.81%</b>	2.71%	2.84%	3.79%	3.11%	1.62%
<b>Chihuahua</b>	Size	175.0	2275.0	2800.0	$\Delta_{Size}$	<b>455.0</b>	-1715.0	-1050.0	490.0	525.0	-980.0
	Acc	53.29%	65.43%	65.0%	$\Delta_{Acc}$	<b>0.29%</b>	1.0%	0.71%	1.0%	0.43%	0.71%
<b>English setter</b>	Size	242.0	3868.0	3881.0	$\Delta_{Size}$	<b>-2573.0</b>	-1323.0	13.0	-588.0	-98.0	-637.0
	Acc	54.69%	66.01%	64.68%	$\Delta_{Acc}$	<b>1.3%</b>	4.63%	1.34%	2.68%	3.4%	2.68%
<b>German police dog</b>	Size	174.0	2379.0	2785.0	$\Delta_{Size}$	<b>35.0</b>	-700.0	-350.0	385.0	406.0	175.0
	Acc	58.25%	65.57%	63.27%	$\Delta_{Acc}$	<b>4.02%</b>	4.73%	4.45%	5.6%	5.6%	7.75%
<b>Standard poodle</b>	Size	195.0	2379.0	3129.0	$\Delta_{Size}$	<b>585.0</b>	-1014.0	-624.0	663.0	750.0	-702.0
	Acc	48.66%	76.63%	73.44%	$\Delta_{Acc}$	<b>1.15%</b>	2.43%	2.04%	<b>1.15%</b>	3.19%	1.53%
<b>Vizsla</b>	Size	233.0	3288.0	3734.0	$\Delta_{Size}$	<b>-517.0</b>	-2021.0	<b>-517.0</b>	282.0	282.0	-846.0
	Acc	60.92%	69.38%	67.24%	$\Delta_{Acc}$	<b>2.03%</b>	3.0%	<b>2.03%</b>	2.36%	2.36%	2.36%
<b>Yorkshire terrier</b>	Size	304.0	4086.0	4874.0	$\Delta_{Size}$	<b>-305.0</b>	-2379.0	-732.0	-183.0	-122.0	61.0
	Acc	49.67%	74.26%	72.87%	$\Delta_{Acc}$	<b>0.33%</b>	1.31%	2.38%	1.97%	1.48%	1.97%
<b>Wild dog</b>	Size	242.0	2546.0	3875.0	$\Delta_{Size}$	1296.0	-912.0	1329.0	1056.0	1152.0	<b>-624.0</b>
	Acc	70.49%	74.82%	73.99%	$\Delta_{Acc}$	<b>0.31%</b>	0.93%	0.83%	0.62%	0.93%	0.52%
<b>Wolf</b>	Size	561.0	4257.0	8986.0	$\Delta_{Size}$	224.0	-2623.0	<b>112.0</b>	1008.0	1568.0	560.0
	Acc	71.66%	76.33%	75.71%	$\Delta_{Acc}$	0.27%	2.44%	0.44%	<b>0.22%</b>	0.4%	2.62%
<b>Fox</b>	Size	478.0	7006.0	7654.0	$\Delta_{Size}$	288.0	-5664.0	-3840.0	<b>-1440.0</b>	-768.0	-2112.0
	Acc	71.94%	76.38%	72.2%	$\Delta_{Acc}$	0.63%	1.52%	2.4%	<b>0.31%</b>	1.04%	1.1%
<b>Rabbit</b>	Size	349.0	4899.0	5593.0	$\Delta_{Size}$	560.0	-2660.0	-1330.0	<b>140.0</b>	694.0	-280.0
	Acc	69.12%	75.98%	73.12%	$\Delta_{Acc}$	<b>1.22%</b>	2.5%	2.93%	3.43%	2.86%	1.5%
<b>Cat</b>	Size	732.0	9405.0	11722.0	$\Delta_{Size}$	<b>-5145.0</b>	-6468.0	-4116.0	1323.0	1617.0	-1470.0
	Acc	61.97%	68.86%	64.5%	$\Delta_{Acc}$	<b>0.78%</b>	1.33%	0.99%	1.19%	2.52%	0.99%
<b>Panda</b>	Size	351.0	3641.0	5628.0	$\Delta_{Size}$	1260.0	-980.0	<b>-630.0</b>	1610.0	1680.0	1987.0
	Acc	76.07%	82.32%	78.98%	$\Delta_{Acc}$	<b>0.21%</b>	1.99%	1.78%	2.84%	2.27%	3.34%
<b>Elephant</b>	Size	392.0	3078.0	6282.0	$\Delta_{Size}$	<b>1699.0</b>	3204.0	3204.0	3204.0	3204.0	3204.0
	Acc	80.66%	85.69%	83.81%	$\Delta_{Acc}$	<b>0.83%</b>	1.88%	1.88%	1.88%	1.88%	1.88%

Table 12. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-LR with PN kernel taking EMC as the strategy. The thresholds of MS are all 2.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Basenji</b>	Size	184.0	2487.0	2951.0	$\Delta_{Size}$	<b>-496.0</b>	464.0	-677.0	464.0	464.0	464.0
	Acc	54.79%	60.98%	58.49%	$\Delta_{Acc}$	<b>1.86%</b>	2.49%	3.52%	2.49%	2.49%	2.49%
<b>Chihuahua</b>	Size	175.0	2435.0	2800.0	$\Delta_{Size}$	<b>-495.0</b>	365.0	-1025.0	365.0	365.0	365.0
	Acc	51.45%	64.33%	62.84%	$\Delta_{Acc}$	<b>1.02%</b>	1.49%	4.41%	1.49%	1.49%	1.49%
<b>English setter</b>	Size	242.0	3390.0	3881.0	$\Delta_{Size}$	<b>-70.0</b>	-297.0	-1225.0	491.0	491.0	491.0
	Acc	54.09%	61.15%	58.14%	$\Delta_{Acc}$	<b>2.01%</b>	5.74%	9.68%	3.01%	3.01%	3.01%
<b>German police dog</b>	Size	174.0	2516.0	2785.0	$\Delta_{Size}$	<b>184.0</b>	270.0	-1458.0	270.0	270.0	270.0
	Acc	56.13%	63.06%	62.34%	$\Delta_{Acc}$	<b>0.47%</b>	0.72%	7.46%	0.72%	0.72%	0.72%
<b>Standard poodle</b>	Size	195.0	2896.0	3129.0	$\Delta_{Size}$	-1782.0	<b>233.0</b>	-1061.0	<b>233.0</b>	<b>233.0</b>	<b>233.0</b>
	Acc	58.01%	69.19%	67.98%	$\Delta_{Acc}$	3.29%	<b>1.21%</b>	1.72%	<b>1.21%</b>	<b>1.21%</b>	<b>1.21%</b>
<b>Vizsla</b>	Size	233.0	2959.0	3734.0	$\Delta_{Size}$	<b>-423.0</b>	775.0	-1285.0	775.0	775.0	775.0
	Acc	53.71%	62.38%	60.85%	$\Delta_{Acc}$	<b>1.46%</b>	1.53%	7.42%	1.53%	1.53%	1.53%
<b>Yorkshire terrier</b>	Size	304.0	4147.0	4874.0	$\Delta_{Size}$	<b>647.0</b>	-1079.0	-1466.0	727.0	727.0	727.0
	Acc	55.05%	71.2%	70.55%	$\Delta_{Acc}$	<b>0.6%</b>	13.09%	9.13%	0.66%	0.66%	0.66%
<b>Wild dog</b>	Size	242.0	2551.0	3875.0	$\Delta_{Size}$	1325.0	1325.0	<b>-404.0</b>	1325.0	1325.0	1325.0
	Acc	62.07%	69.79%	69.17%	$\Delta_{Acc}$	<b>0.62%</b>	<b>0.62%</b>	2.37%	<b>0.62%</b>	<b>0.62%</b>	<b>0.62%</b>
<b>Wolf</b>	Size	561.0	8743.0	8986.0	$\Delta_{Size}$	-5690.0	<b>243.0</b>	<b>243.0</b>	<b>243.0</b>	<b>243.0</b>	<b>243.0</b>
	Acc	62.53%	70.2%	69.53%	$\Delta_{Acc}$	2.32%	<b>0.67%</b>	<b>0.67%</b>	<b>0.67%</b>	<b>0.67%</b>	<b>0.67%</b>
<b>Fox</b>	Size	478.0	4606.0	7654.0	$\Delta_{Size}$	<b>1680.0</b>	3048.0	3048.0	3048.0	3048.0	3048.0
	Acc	63.22%	70.9%	69.17%	$\Delta_{Acc}$	<b>1.02%</b>	1.72%	1.72%	1.72%	1.72%	1.72%
<b>Rabbit</b>	Size	349.0	3359.0	5593.0	$\Delta_{Size}$	<b>-490.0</b>	2234.0	2234.0	2234.0	2234.0	2234.0
	Acc	56.15%	68.87%	66.01%	$\Delta_{Acc}$	<b>2.68%</b>	2.86%	2.86%	2.86%	2.86%	2.86%
<b>Cat</b>	Size	732.0	1908.0	11722.0	$\Delta_{Size}$	<b>-294.0</b>	9814.0	9814.0	9814.0	9814.0	9814.0
	Acc	64.36%	65.11%	63.4%	$\Delta_{Acc}$	<b>1.26%</b>	1.71%	1.71%	1.71%	1.71%	1.71%
<b>Panda</b>	Size	351.0	5563.0	5628.0	$\Delta_{Size}$	<b>47.0</b>	65.0	-901.0	65.0	65.0	65.0
	Acc	73.11%	78.5%	78.39%	$\Delta_{Acc}$	<b>0.09%</b>	0.12%	5.23%	0.12%	0.12%	0.12%
<b>Elephant</b>	Size	392.0	5646.0	6282.0	$\Delta_{Size}$	<b>338.0</b>	637.0	-1817.0	637.0	637.0	637.0
	Acc	73.12%	78.91%	77.23%	$\Delta_{Acc}$	<b>1.59%</b>	1.69%	16.03%	1.69%	1.69%	1.69%

**A.2.2 Experiments for Kernel SVM.** Table 13-Table 14 present the results of kernel SVM on UCI data sets with RBF kernel and PN kernel, respectively. Table 15-Table 16 show the results of kernel SVM on ImageNet data sets. As before, MS is still observed to outperform the baselines and it achieves the highest accuracy with fewer training examples on most cases.

Table 13. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-SVM with RBF kernel taking EMC as the strategy. The thresholds of MS are 2.0, 1.0, 1.5, 2.5, 2.0, 2.0, 2.0, 2.0, 1.0 respectively.

Dataset	Initial	HPP	All	MS	SP	GRAD	OU	MC	ME	QUIRE
<b>Biodeg</b>	Size 52.0	320.0	843.0	$\Delta_{Size}$ <b>7.0</b>	97.0	124.0	523.0	523.0	142.0	151.0
	Acc 77.83%	88.38%	87.19%	$\Delta_{Acc}$ <b>0.99%</b>	1.56%	2.04%	1.19%	1.19%	1.51%	2.93%
<b>Clave</b>	Size 430.0	5031.0	6880.0	$\Delta_{Size}$ <b>-2795.0</b>	-3712.0	-3698.0	1849.0	1849.0	-4601.0	-1063.0
	Acc 96.2%	98.31%	98.17%	$\Delta_{Acc}$ <b>0.13%</b>	0.39%	0.21%	0.15%	0.15%	2.11%	0.25%
<b>Ionosphere</b>	Size 17.0	75.0	280.0	$\Delta_{Size}$ 50.0	55.0	76.0	200.0	205.0	198.0	<b>34.0</b>
	Acc 76.46%	94.97%	93.56%	$\Delta_{Acc}$ <b>1.01%</b>	1.21%	2.21%	1.41%	1.41%	1.21%	10.26%
<b>Spambase</b>	Size 230.0	1741.0	3680.0	$\Delta_{Size}$ <b>-434.0</b>	-710.0	486.0	1939.0	1939.0	1209.0	-355.0
	Acc 87.36%	93.62%	93.36%	$\Delta_{Acc}$ <b>0.17%</b>	0.43%	0.51%	0.26%	0.26%	0.26%	1.49%
<b>WDBC</b>	Size 28.0	149.0	454.0	$\Delta_{Size}$ 302.0	-40.0	71.0	<b>30.0</b>	305.0	51.0	-70.0
	Acc 93.39%	98.0%	97.3%	$\Delta_{Acc}$ <b>0.7%</b>	1.04%	0.83%	0.91%	<b>0.7%</b>	0.96%	2.35%
<b>D-vs-P</b>	Size 80.0	185.0	1286.0	$\Delta_{Size}$ <b>-11.0</b>	-18.0	247.0	9.0	1086.0	-97.0	175.0
	Acc 97.12%	99.38%	99.21%	$\Delta_{Acc}$ <b>0.17%</b>	0.19%	0.19%	0.19%	<b>0.17%</b>	1.95%	0.23%
<b>E-vs-F</b>	Size 77.0	294.0	1234.0	$\Delta_{Size}$ <b>-85.0</b>	-93.0	119.0	677.0	939.0	940.0	23.0
	Acc 95.1%	99.55%	99.14%	$\Delta_{Acc}$ <b>0.3%</b>	0.39%	0.32%	0.43%	0.37%	0.41%	0.8%
<b>M-vs-N</b>	Size 78.0	906.0	1259.0	$\Delta_{Size}$ <b>-93.0</b>	-709.0	-562.0	353.0	353.0	353.0	-607.0
	Acc 94.43%	98.99%	98.54%	$\Delta_{Acc}$ <b>0.38%</b>	1.84%	1.2%	0.44%	0.44%	0.44%	1.58%
<b>U-vs-V</b>	Size 78.0	259.0	1260.0	$\Delta_{Size}$ 420.0	<b>-82.0</b>	149.0	548.0	1001.0	-181.0	580.0
	Acc 97.41%	99.65%	99.54%	$\Delta_{Acc}$ <b>0.07%</b>	1.54%	0.7%	0.11%	0.11%	2.24%	0.11%

Table 14. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on UCI data sets for kernel-SVM with PN kernel taking EMC as the strategy. The thresholds of MS are 1.5, 1.5, 0.5, 1.5, 2.0, 1.5, 1.5, 2.0, 1.5 respectively.

Dataset	Initial	HPP	All	MS	SP	GRAD	OU	MC	ME	QUIRE
<b>Biodeg</b>	Size 52.0	433.0	843.0	$\Delta_{Size}$ <b>-103.0</b>	-42.0	-166.0	410.0	410.0	410.0	-181.0
	Acc 78.9%	89.34%	87.64%	$\Delta_{Acc}$ <b>1.07%</b>	1.26%	2.83%	1.7%	1.7%	1.7%	3.27%
<b>Clave</b>	Size 430.0	1763.0	6880.0	$\Delta_{Size}$ <b>-275.0</b>	-512.0	-353.0	4390.0	5117.0	-1333.0	3861.0
	Acc 94.9%	97.67%	96.34%	$\Delta_{Acc}$ <b>0.24%</b>	0.31%	0.29%	1.21%	1.33%	2.77%	1.1%
<b>Ionosphere</b>	Size 17.0	142.0	280.0	$\Delta_{Size}$ <b>-17.0</b>	-10.0	-2.0	138.0	138.0	116.0	-89.0
	Acc 74.65%	91.55%	87.14%	$\Delta_{Acc}$ <b>3.57%</b>	4.98%	5.35%	4.41%	4.41%	4.79%	9.2%
<b>Spambase</b>	Size 230.0	2290.0	3680.0	$\Delta_{Size}$ -1126.0	-1314.0	<b>106.0</b>	1089.0	1390.0	775.0	-1603.0
	Acc 86.77%	93.0%	92.84%	$\Delta_{Acc}$ 1.19%	1.8%	1.06%	<b>0.15%</b>	0.16%	0.25%	3.18%
<b>WDBC</b>	Size 28.0	109.0	454.0	$\Delta_{Size}$ <b>-49.0</b>	-22.0	83.0	285.0	345.0	307.0	-47.0
	Acc 92.61%	98.35%	97.13%	$\Delta_{Acc}$ <b>0.87%</b>	1.3%	1.39%	1.22%	1.22%	1.3%	1.65%
<b>D-vs-P</b>	Size 80.0	178.0	1286.0	$\Delta_{Size}$ <b>-9.0</b>	-11.0	286.0	30.0	1099.0	28.0	377.0
	Acc 96.19%	99.36%	99.13%	$\Delta_{Acc}$ <b>0.12%</b>	0.21%	0.25%	0.19%	0.23%	0.19%	0.19%
<b>E-vs-F</b>	Size 77.0	181.0	1234.0	$\Delta_{Size}$ <b>-2.0</b>	<b>2.0</b>	176.0	320.0	1050.0	1054.0	215.0
	Acc 96.21%	98.96%	98.77%	$\Delta_{Acc}$ 0.45%	0.26%	0.16%	0.16%	0.19%	0.19%	<b>0.13%</b>
<b>M-vs-N</b>	Size 78.0	246.0	1259.0	$\Delta_{Size}$ <b>-16.0</b>	-47.0	234.0	529.0	996.0	1013.0	172.0
	Acc 95.39%	99.59%	99.25%	$\Delta_{Acc}$ <b>0.25%</b>	0.41%	0.36%	0.29%	0.36%	0.34%	0.38%
<b>U-vs-V</b>	Size 78.0	184.0	1260.0	$\Delta_{Size}$ <b>-14.0</b>	-35.0	179.0	513.0	1012.0	-106.0	536.0
	Acc 97.6%	99.87%	99.68%	$\Delta_{Acc}$ <b>0.09%</b>	0.32%	0.19%	0.19%	0.16%	2.27%	0.19%

Table 15. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-SVM with RBF kernel taking EMC as the strategy. The thresholds of MS are 1.0, 1.0, 1.0, 1.5, 1.5, 1.0, 1.0, 1.0, 0.5, 0.5, 1.0, 1.0, 0.5, 1.0 respectively.

Dataset	Initial	HPP	All	MS	SP	GRAD	OU	MC	ME	QUIRE
<b>Basenji</b>	<i>Size</i> 184.0	2355.0	2951.0	$\Delta_{Size}$ <b>596.0</b>	<b>596.0</b>	-999.0	<b>596.0</b>	<b>596.0</b>	<b>596.0</b>	-987.0
	<i>Acc</i> 51.83%	66.67%	65.76%	$\Delta_{Acc}$ <b>0.9%</b>	0.9%	3.61%	<b>0.9%</b>	<b>0.9%</b>	<b>0.9%</b>	3.25%
<b>Chihuahua</b>	<i>Size</i> 175.0	2188.0	2800.0	$\Delta_{Size}$ <b>284.0</b>	346.0	-324.0	613.0	613.0	613.0	-1330.0
	<i>Acc</i> 58.36%	65.2%	64.3%	$\Delta_{Acc}$ <b>0.88%</b>	0.93%	1.73%	0.89%	0.89%	0.89%	2.98%
<b>English setter</b>	<i>Size</i> 242.0	2982.0	3881.0	$\Delta_{Size}$ <b>702.0</b>	857.0	-984.0	899.0	899.0	899.0	-1594.0
	<i>Acc</i> 53.53%	65.18%	64.48%	$\Delta_{Acc}$ <b>0.59%</b>	0.71%	2.82%	0.7%	0.7%	0.7%	3.33%
<b>German police dog</b>	<i>Size</i> 174.0	1812.0	2785.0	$\Delta_{Size}$ 655.0	973.0	<b>-122.0</b>	973.0	973.0	973.0	-1057.0
	<i>Acc</i> 56.47%	63.87%	62.96%	$\Delta_{Acc}$ <b>0.83%</b>	0.92%	3.85%	0.92%	0.92%	0.92%	3.79%
<b>Standard poodle</b>	<i>Size</i> 195.0	2353.0	3129.0	$\Delta_{Size}$ <b>149.0</b>	217.0	-796.0	776.0	776.0	260.0	-1564.0
	<i>Acc</i> 65.9%	74.64%	74.02%	$\Delta_{Acc}$ <b>0.38%</b>	0.68%	2.31%	0.62%	0.62%	0.74%	4.37%
<b>Vizsla</b>	<i>Size</i> 233.0	3153.0	3734.0	$\Delta_{Size}$ 581.0	<b>171.0</b>	-611.0	581.0	581.0	581.0	-2027.0
	<i>Acc</i> 59.64%	68.52%	68.31%	$\Delta_{Acc}$ <b>0.21%</b>	0.43%	1.63%	<b>0.21%</b>	<b>0.21%</b>	<b>0.21%</b>	3.1%
<b>Yorkshire terrier</b>	<i>Size</i> 304.0	4140.0	4874.0	$\Delta_{Size}$ <b>-349.0</b>	-609.0	-398.0	734.0	734.0	-622.0	-2348.0
	<i>Acc</i> 68.51%	74.84%	74.51%	$\Delta_{Acc}$ 0.65%	0.66%	0.56%	<b>0.33%</b>	<b>0.33%</b>	0.76%	2.11%
<b>Wild dog</b>	<i>Size</i> 242.0	2919.0	3875.0	$\Delta_{Size}$ 760.0	<b>-304.0</b>	-713.0	956.0	956.0	956.0	-1589.0
	<i>Acc</i> 66.86%	73.93%	73.44%	$\Delta_{Acc}$ <b>0.45%</b>	0.86%	1.22%	0.48%	0.48%	0.48%	2.08%
<b>Wolf</b>	<i>Size</i> 561.0	8177.0	8986.0	$\Delta_{Size}$ <b>809.0</b>	-2688.0	<b>809.0</b>	<b>809.0</b>	809.0	-2725.0	-4256.0
	<i>Acc</i> 71.71%	77.21%	77.09%	$\Delta_{Acc}$ <b>0.12%</b>	0.55%	<b>0.12%</b>	<b>0.12%</b>	<b>0.12%</b>	0.53%	1.5%
<b>Fox</b>	<i>Size</i> 478.0	7195.0	7654.0	$\Delta_{Size}$ <b>419.0</b>	-2408.0	459.0	459.0	459.0	-2227.0	-1885.0
	<i>Acc</i> 71.26%	77.75%	77.59%	$\Delta_{Acc}$ <b>0.15%</b>	0.73%	0.16%	0.16%	0.16%	0.67%	0.72%
<b>Rabbit</b>	<i>Size</i> 349.0	4598.0	5593.0	$\Delta_{Size}$ -693.0	-574.0	<b>226.0</b>	995.0	995.0	-595.0	-1897.0
	<i>Acc</i> 68.43%	76.48%	76.21%	$\Delta_{Acc}$ 0.72%	0.64%	0.62%	<b>0.26%</b>	<b>0.26%</b>	0.63%	1.73%
<b>Cat</b>	<i>Size</i> 732.0	10022.0	11722.0	$\Delta_{Size}$ 1700.0	-1852.0	<b>-1682.0</b>	1700.0	1700.0	1700.0	-1154.0
	<i>Acc</i> 64.68%	69.28%	68.92%	$\Delta_{Acc}$ <b>0.36%</b>	0.49%	1.12%	<b>0.36%</b>	<b>0.36%</b>	<b>0.36%</b>	0.67%
<b>Panda</b>	<i>Size</i> 351.0	4341.0	5628.0	$\Delta_{Size}$ <b>-1400.0</b>	-1372.0	1287.0	1287.0	1287.0	764.0	-1451.0
	<i>Acc</i> 77.64%	83.17%	82.77%	$\Delta_{Acc}$ <b>0.34%</b>	0.37%	0.4%	0.4%	0.4%	0.38%	0.8%
<b>Elephant</b>	<i>Size</i> 392.0	4387.0	6282.0	$\Delta_{Size}$ <b>-818.0</b>	-1151.0	1510.0	1895.0	1895.0	-450.0	-2054.0
	<i>Acc</i> 79.94%	85.23%	84.86%	$\Delta_{Acc}$ <b>0.37%</b>	0.48%	0.42%	<b>0.37%</b>	<b>0.37%</b>	0.52%	0.85%

Table 16. The results of  $\Delta_{Size}$  and  $\Delta_{Acc}$  on ImageNet data sets for kernel-SVM with PN kernel taking EMC as the strategy. The thresholds of MS are all 1.0.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME	QUIRE
Basenji	Size	184.0	2490.0	2951.0	$\Delta_{Size}$	<b>407.0</b>	461.0	-2060.0	461.0	461.0	461.0	-691.0
	Acc	54.8%	65.36%	64.37%	$\Delta_{Acc}$	<b>0.81%</b>	0.99%	4.51%	0.99%	0.99%	0.99%	0.9%
Chihuahua	Size	175.0	2076.0	2800.0	$\Delta_{Size}$	<b>518.0</b>	725.0	-1173.0	725.0	725.0	725.0	-1022.0
	Acc	56.09%	66.41%	64.73%	$\Delta_{Acc}$	<b>1.41%</b>	1.69%	4.63%	1.69%	1.69%	1.69%	3.4%
English setter	Size	242.0	3011.0	3881.0	$\Delta_{Size}$	<b>821.0</b>	871.0	-1493.0	871.0	871.0	871.0	-1795.0
	Acc	53.22%	64.8%	63.03%	$\Delta_{Acc}$	<b>1.54%</b>	1.78%	3.51%	1.78%	1.78%	1.78%	3.58%
German police dog	Size	174.0	2527.0	2785.0	$\Delta_{Size}$	<b>153.0</b>	258.0	-1320.0	258.0	258.0	258.0	-1310.0
	Acc	55.84%	65.08%	64.16%	$\Delta_{Acc}$	<b>0.53%</b>	0.93%	3.68%	0.93%	0.93%	0.93%	3.16%
Standard poodle	Size	195.0	2535.0	3129.0	$\Delta_{Size}$	94.0	<b>-82.0</b>	-1435.0	594.0	594.0	594.0	-1377.0
	Acc	61.44%	75.71%	74.75%	$\Delta_{Acc}$	<b>0.64%</b>	0.79%	5.24%	0.96%	0.96%	0.96%	3.68%
Vizsla	Size	233.0	3250.0	3734.0	$\Delta_{Size}$	<b>362.0</b>	484.0	-1612.0	484.0	484.0	484.0	-1166.0
	Acc	59.59%	70.84%	70.02%	$\Delta_{Acc}$	<b>0.71%</b>	0.81%	3.9%	0.81%	0.81%	0.81%	3.1%
Yorkshire terrier	Size	304.0	3647.0	4874.0	$\Delta_{Size}$	775.0	-360.0	<b>-95.0</b>	1227.0	1227.0	1216.0	-1354.0
	Acc	68.63%	76.11%	75.5%	$\Delta_{Acc}$	<b>0.56%</b>	0.87%	1.49%	0.61%	0.61%	0.66%	1.48%
Wild dog	Size	242.0	3015.0	3875.0	$\Delta_{Size}$	69.0	<b>50.0</b>	-1389.0	860.0	860.0	860.0	-1074.0
	Acc	63.92%	74.96%	74.39%	$\Delta_{Acc}$	1.82%	1.11%	3.38%	<b>0.58%</b>	<b>0.58%</b>	<b>0.58%</b>	1.98%
Wolf	Size	561.0	4369.0	8986.0	$\Delta_{Size}$	4480.0	1120.0	4617.0	4617.0	4617.0	4617.0	<b>224.0</b>
	Acc	71.89%	76.89%	76.53%	$\Delta_{Acc}$	<b>0.33%</b>	0.42%	0.36%	0.36%	0.36%	0.36%	0.58%
Fox	Size	478.0	7294.0	7654.0	$\Delta_{Size}$	<b>-480.0</b>	-2400.0	360.0	360.0	360.0	360.0	-2304.0
	Acc	71.68%	77.59%	77.27%	$\Delta_{Acc}$	<b>0.29%</b>	0.78%	0.31%	0.31%	0.31%	0.31%	0.76%
Rabbit	Size	349.0	4787.0	5593.0	$\Delta_{Size}$	<b>420.0</b>	-588.0	-802.0	806.0	806.0	806.0	-1344.0
	Acc	69.09%	76.7%	76.23%	$\Delta_{Acc}$	<b>0.46%</b>	0.86%	1.52%	0.47%	0.47%	0.47%	0.84%
Cat	Size	732.0	8964.0	11722.0	$\Delta_{Size}$	2702.0	882.0	-1635.0	2758.0	2758.0	2758.0	<b>-221.0</b>
	Acc	64.51%	70.0%	69.65%	$\Delta_{Acc}$	0.32%	<b>0.29%</b>	1.48%	0.36%	0.36%	0.36%	0.36%
Panda	Size	351.0	3431.0	5628.0	$\Delta_{Size}$	1081.0	-809.0	<b>269.0</b>	2197.0	2197.0	1784.0	-1003.0
	Acc	76.63%	84.19%	83.66%	$\Delta_{Acc}$	<b>0.5%</b>	0.52%	1.18%	0.53%	0.53%	0.51%	0.77%
Elephant	Size	392.0	4842.0	6282.0	$\Delta_{Size}$	<b>527.0</b>	-2159.0	1440.0	1440.0	1440.0	-2107.0	-2370.0
	Acc	78.94%	85.43%	84.84%	$\Delta_{Acc}$	<b>0.49%</b>	1.1%	0.59%	0.59%	0.59%	1.21%	1.21%

## B EXPERIMENTS FOR NON-GRADIENT-BASED MODELS

In this section, we validate our proposed algorithm on one non-gradient-based model: the gaussian process regression (GPR) model (Section 7.2). We conduct experiments on four regression data sets: **Concrete**, **Housing**, **WineRed** from the UCI machine learning repository and **PM10** from the StatLib repository<sup>1</sup>. The performance of the regression model is measured by the root mean square error (RMSE) [1]. Similarly, we use  $\Delta_{Size}$  and  $\Delta_{Rmse} = Rmse_{SC} - Rmse_{HPP}$  to evaluate the performance of each stopping criterion.

Table 17 presents the comparison results. Our proposed MS is observed to perform the best and it obtains the lowest RMSE with fewer annotation costs, which demonstrates the effectiveness of our method on GPR model. The other methods perform not well which is similar to previous results.

Table 17. The results of  $\Delta_{Size}$  and  $\Delta_{Rmse}$  on UCI and StatLib data sets for GPR taking EMC as the active learning method. The thresholds of MS are all 0.1.

Dataset		Initial	HPP	All		MS	SP	GRAD	OU	MC	ME
<b>Concrete</b>	<i>Size</i>	51.0	746.0	823.0	$\Delta_{Size}$	<b>15.0</b>	-95.0	77.0	55.0	75.0	77.0
	<i>Rmse</i>	13.4535	5.4988	5.6037	$\Delta_{Rmse}$	<b>0.0347</b>	1.1227	0.1049	0.0989	0.1044	0.1049
<b>Housing</b>	<i>Size</i>	25.0	333.0	404.0	$\Delta_{Size}$	<b>6.0</b>	-96.0	37.0	-8.0	69.0	71.0
	<i>Rmse</i>	7.8891	3.2776	3.3266	$\Delta_{Rmse}$	<b>0.0192</b>	1.4685	0.9982	0.7145	0.0493	0.0489
<b>PM10</b>	<i>Size</i>	25.0	334.0	400.0	$\Delta_{Size}$	<b>6.0</b>	-273.0	66.0	66.0	66.0	66.0
	<i>Rmse</i>	0.9443	0.8061	0.8078	$\Delta_{Rmse}$	<b>0.0011</b>	0.065	0.0017	0.0017	0.0017	0.0017
<b>WineRed</b>	<i>Size</i>	79.0	963.0	1278.0	$\Delta_{Size}$	204.0	-830.0	315.0	-884.0	-820.0	<b>36.0</b>
	<i>Rmse</i>	0.7987	0.6363	0.6386	$\Delta_{Rmse}$	<b>0.0018</b>	0.2018	0.0023	0.1623	0.1521	0.0026

## REFERENCES

- [1] W. Cai, M. Zhang, and Y. Zhang. 2016. Batch mode active learning for regression with expected model change. *IEEE Transactions on Neural Networks and Learning Systems* (2016).

<sup>1</sup><http://lib.stat.cmu.edu/datasets/>