

# Multi-Flow:

## Multi-View-Enriched Normalizing Flows for Industrial Anomaly Detection

### Supplementary Material

#### 6. Comprehensive Detection Results

We provide the full results for detection of all baselines on all classes of Real-IAD in Tab. 5 and Tab. 6.

Classes	PaDiM [12]	Cflow [21]	SoftPatch [26]	DeSTSeg [64]	RD [13]	UniAD [58]	PatchCore [48]	SimpleNet [38]	<i>Multi-Flow</i> (ours)
Audiojack	92.2	82.0	91.0	95.3	81.9	91.2	89.3	91.2	<b>96.56</b> ± 0.12
Bottle Cap	98.1	98.8	99.1	92.4	93.7	97.3	<b>99.4</b>	<b>99.4</b>	98.00 ± 0.14
Button Battery	88.7	96.3	91.9	93.3	83.3	87.5	90.6	95.8	<b>96.57</b> ± 0.16
End Cap	76.1	75.1	<u>92.3</u>	82.3	68.1	89.4	91.9	<b>94.2</b>	90.89 ± 0.09
Eraser	<b>96.5</b>	87.2	96.1	91.9	82.9	91.2	<u>95.6</u>	94.7	92.81 ± 0.14
Fire Hood	<b>96.9</b>	87	87.8	<b>96.9</b>	81.4	83.0	89.3	95.6	95.45 ± 0.10
Mint	69.1	79.1	82.1	<u>77.7</u>	67.7	73.0	85.7	<u>86.8</u>	<b>87.84</b> ± 0.03
Mounts	98.4	98.2	99.3	99.1	92.5	97.0	<b>99.7</b>	99.4	<u>99.46</u> ± 0.02
PCB	88.4	83.1	90.3	83.6	79.3	83.2	<u>93</u>	90.7	<b>93.91</b> ± 0.11
Phone Battery	91.7	91.2	91.5	<u>98.2</u>	89.4	93.6	95.1	94.7	<b>98.68</b> ± 0.06
Plastic Nut	<b>98.2</b>	88.6	95.7	94.4	72.8	87.1	97.8	95.7	<b>98.96</b> ± 0.10
Plastic Plug	87.4	90.0	92.5	<u>95.6</u>	89.3	78.0	<b>95.7</b>	94.4	93.85 ± 0.07
Porcelain Doll	93.8	<u>95.1</u>	94.7	94.6	89.6	92.8	96.1	<b>96.2</b>	94.37 ± 0.42
Regulator	<b>96.5</b>	85.1	82.9	93.0	92.5	55.5	86.0	92.0	<u>94.22</u> ± 0.22
Rolled Strip Base	98.6	98.8	99.7	98.9	80.3	99.3	<b>99.7</b>	<u>99.6</u>	<u>99.57</u> ± 0.09
SIM Card Set	94.2	95.6	98.4	98.3	89.9	94.0	<b>99.3</b>	<u>99.2</u>	98.79 ± 0.04
Switch	82.1	92.9	97.8	96.6	87.3	95.3	94.6	<u>98.8</u>	<b>99.27</b> ± 0.03
Tape	99.8	98.5	99.7	99.1	89.5	99.1	<u>99.9</u>	<b>100</b>	97.12 ± 0.12
Terminal Block	96.9	92.2	<u>98.2</u>	96.1	89.8	93.8	97.5	97.7	<b>99.09</b> ± 0.02
Toothbrush	91.7	91.9	92.9	<b>97.9</b>	86.7	95.0	94.7	95.3	96.88 ± 0.13
Toy	91.4	78.8	91.3	<b>96.5</b>	75.0	77.2	92.8	92.9	<u>96.46</u> ± 0.19
Toy Brick	84.3	82.9	78.2	<u>87</u>	72.5	78.3	82.6	85.7	<b>91.73</b> ± 0.14
Transistor1	90.3	96.6	99.3	99.0	94.7	99.3	<b>99.8</b>	<u>99.7</u>	98.55 ± 0.07
U Block	98.3	96.7	98.3	<u>98.5</u>	86.9	96.3	<b>98.8</b>	<u>98.5</u>	96.57 ± 0.10
USB	77.0	86.1	93.8	93.3	89.4	83.1	<u>93.9</u>	<u>93.9</u>	<b>96.83</b> ± 0.04
USB Adaptor	93.2	86.8	91.9	<b>93.6</b>	65.3	85.1	<u>90.6</u>	93.0	<u>93.39</u> ± 0.18
Vcpill	94.7	87.8	93.7	96.4	87.2	89.4	<u>96.5</u>	<b>97.5</b>	95.90 ± 0.11
Wooden Beads	91.1	89.3	90.9	91.9	85	82.5	91.4	<u>92.9</u>	<b>94.72</b> ± 0.16
Woodstick	81.8	83.9	73.9	<b>90.2</b>	71.9	76.0	74.5	81.5	<u>89.15</u> ± 0.19
Zipper	99.3	97.6	99.6	99.7	96.1	98.8	<b>100</b>	99.7	<u>99.92</u> ± 0.00
Average	91.2	89.8	92.8	94.0	83.7	88.1	93.7	<u>94.9</u>	<b>95.85</b> ± 0.02

Table 5. Performance in sample-wise anomaly detection on all classes of Real-IAD [54]. Here, the anomaly scores for all images belonging to the same object are aggregated into one value. Best performances are **bold**, with the runner-up underlined. We execute our method for  $n = 5$  runs and report mean ± standard deviation.

Classes	PaDiM [12]	Cflow [21]	SoftPatch [26]	DeSTSeg [64]	RD [13]	UniAD [58]	PatchCore [48]	SimpleNet [38]	<i>Multi-Flow</i> (ours)
Audiojack	66.6	74.3	<u>88.5</u>	81.8	82.4	82.8	86.3	87.4	<b>90.69</b> ± 0.06
Bottle Cap	87.7	91.3	<b>95.9</b>	87.1	89.2	89.8	94.3	91.6	<u>94.62</u> ± 0.14
Button Battery	84.8	85.2	88.5	<b>91.2</b>	87.0	79	87.3	88.8	<u>89.37</u> ± 0.16
End Cap	73.7	66.4	<b>85.8</b>	80.3	79.0	80.4	84.1	83.4	<u>84.77</u> ± 0.10
Eraser	86.7	88.1	<b>93.5</b>	88.2	89.2	89.6	<u>93.4</u>	91.2	92.08 ± 0.11
Fire Hood	77.3	80.5	<u>84.3</u>	78.9	<u>84.3</u>	79.5	84.1	81.8	<b>93.35</b> ± 0.13
Mint	66.9	70.7	74.5	70.5	<u>71.6</u>	67.6	76.2	<u>77.2</u>	<b>79.38</b> ± 0.11
Mounts	82.5	85.3	85.9	85.1	85.7	87.2	<u>88</u>	<b>88.2</b>	84.26 ± 0.11
PCB	75.0	77.3	90.8	79.6	<u>89.5</u>	80.5	<b>92.4</b>	87.1	<u>91.79</u> ± 0.09
Phone Battery	81.9	84.4	90.2	86.2	<u>90.8</u>	83.4	<b>91.6</b>	88.9	<u>90.77</u> ± 0.35
Plastic Nut	73.8	79.8	89.3	<u>90.2</u>	<u>85.0</u>	79.3	<b>90.8</b>	89.3	88.04 ± 0.13
Plastic Plug	80.0	83.9	88.7	<u>86.0</u>	<b>90.5</b>	78.2	<u>89.7</u>	87.1	88.45 ± 0.11
Porcelain Doll	74.3	76.0	86.1	83.7	<u>87.8</u>	84.1	<b>88.2</b>	86.1	86.62 ± 0.23
Regulator	76.5	62.9	82.1	<b>89.8</b>	<u>87.3</u>	51.8	81.9	82.2	<u>87.69</u> ± 0.18
Rolled Strip Base	97.4	97.1	<u>99.1</u>	98.3	94.3	98.6	98.9	<b>99.4</b>	98.15 ± 0.12
SIM Card Set	91.8	94.4	<u>95.8</u>	91.8	93.6	91.1	<b>97.1</b>	<u>96.1</u>	90.65 ± 0.16
Switch	80.8	83.9	94.7	92.5	<u>87.2</u>	85.7	89.4	<u>94.8</u>	<b>96.03</b> ± 0.12
Tape	93.7	96.0	<u>97.9</u>	96.1	93.1	97.2	<b>98.2</b>	96.9	94.32 ± 0.11
Terminal Block	86.7	85.1	<u>96.2</u>	91.4	94.5	85.9	94.1	93.4	<b>96.41</b> ± 0.19
Toothbrush	74.4	79.9	<b>89.3</b>	87.5	83.4	82.7	<u>88.0</u>	86.1	85.21 ± 0.17
Toy	82.5	67.8	86.7	82.1	81.5	67.3	<u>87.3</u>	82.2	<b>88.31</b> ± 0.10
Toy Brick	71.0	80.7	79.8	76.8	73.9	77.9	<u>81.4</u>	80.8	<b>87.35</b> ± 0.01
Transistor1	86.5	92.8	97.2	95.6	95.8	94.1	<b>97.8</b>	<u>97.3</u>	94.86 ± 0.14
U Block	84.1	90.5	91.6	90.1	<u>91.7</u>	88.8	<b>91.9</b>	90.7	90.45 ± 0.19
USB	69.7	78.1	91.2	90.3	<u>92.1</u>	80.9	90.7	90.0	<b>93.37</b> ± 0.03
USB Adaptor	81.9	75.2	<u>86</u>	73.1	71.9	77.3	83.8	82.7	<b>88.07</b> ± 0.13
Vcpill	67.4	85.8	<u>89.6</u>	90.0	90.2	88.6	<b>91.4</b>	<u>91.1</u>	89.50 ± 0.19
Wooden Beads	82.4	87.4	<b>89.2</b>	86.2	87.4	80.7	<u>88.6</u>	85.7	87.28 ± 0.05
Woodstick	79.9	78.9	76.0	<b>89</b>	84.2	78.9	<u>77.2</u>	77.9	88.04 ± 0.17
Zipper	90.4	93.8	97.9	98.4	97.8	98.2	98.1	<u>98.2</u>	<b>98.23</b> ± 0.08
Average	80.3	82.5	<u>89.4</u>	86.9	87.1	82.9	<u>89.4</u>	88.5	<b>90.27</b> ± 0.02

Table 6. Performance in image-wise anomaly detection on all classes of Real-IAD [54]. Here, the anomaly scores for all images belonging to the same object are aggregated into one value. Best performances are **bold**, with the runner-up underlined. We execute our method for  $n = 5$  runs and report mean ± standard deviation.

Metrics	PaDiM [12]	Cflow [21]	SoftPatch [26]	DeSTSeg [64]	RD [13]	UniAD [58]	PatchCore [48]	SimpleNet [38]	<i>Multi-Flow</i> (ours)
AUPRO	81.8	90.5	90.8	81.5	<b>93.8</b>	86.1	<u>91.5</u>	84.6	87.91 ± 0.01
Pixel-AUROC	-	-	-	-	-	-	-	-	96.47 ± 0.00

Table 7. **Average** segmentation performance of various methods across all classes in Real-IAD [54]. Pixel-wise AUROC has not been reported for the baselines. Best performances are **bold**, with the runner-up underlined. We execute our method for  $n = 5$  runs and report mean ± standard deviation.