

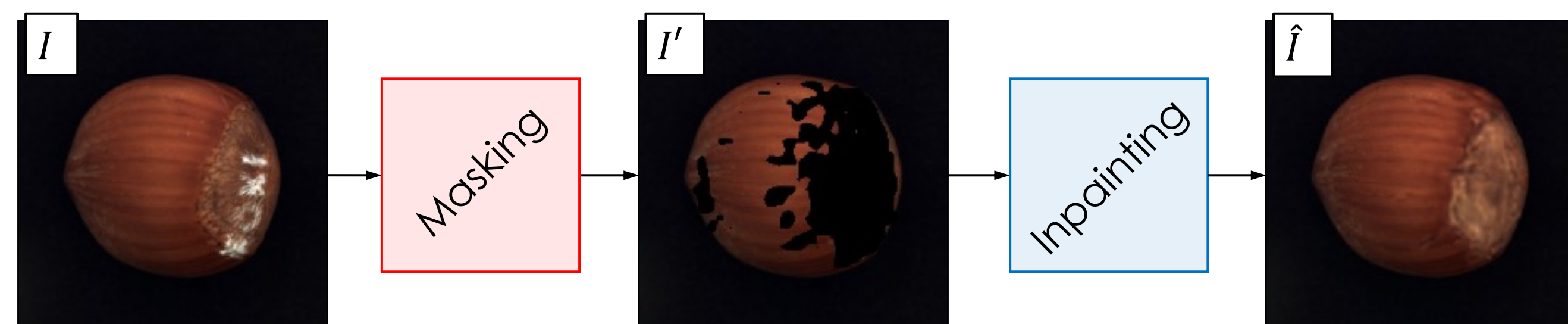
Feature Attenuation of Defective Representation Can Resolve Incomplete Masking on Anomaly Detection

YeongHyeon Park ^{1,2} Sungho Kang ¹ Myung Jin Kim ² Hyeong Seok Kim ² Juneho Yi ¹
¹Sungkyunkwan University. ²SK Planet Co., Ltd.



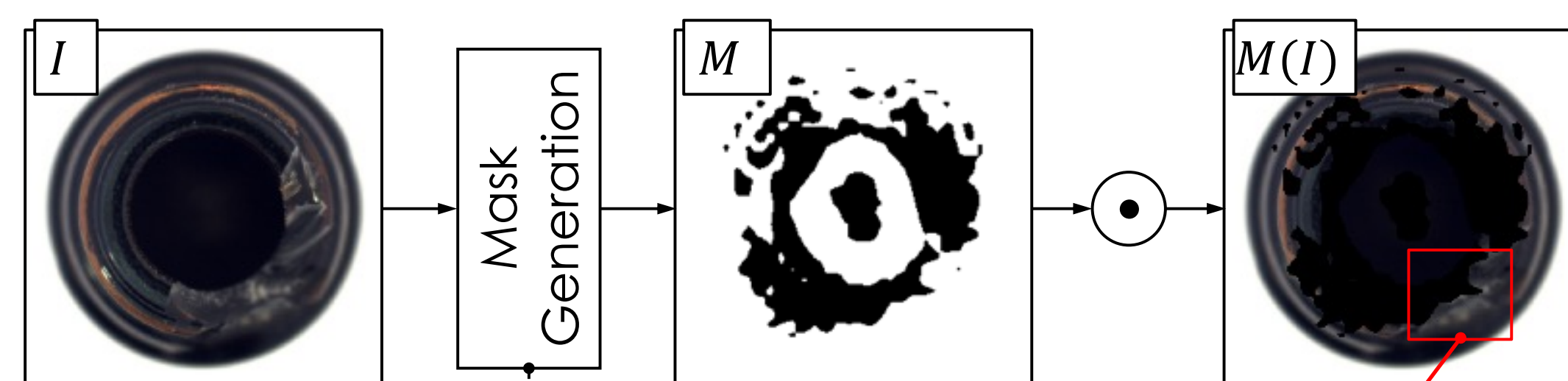
Reconstruction-by-inpainting

- Mask out the suspicious anomalous region
- Determine abnormality based on inpainting error, $\mathcal{L}(I, \hat{I})$



Motivation & Solution

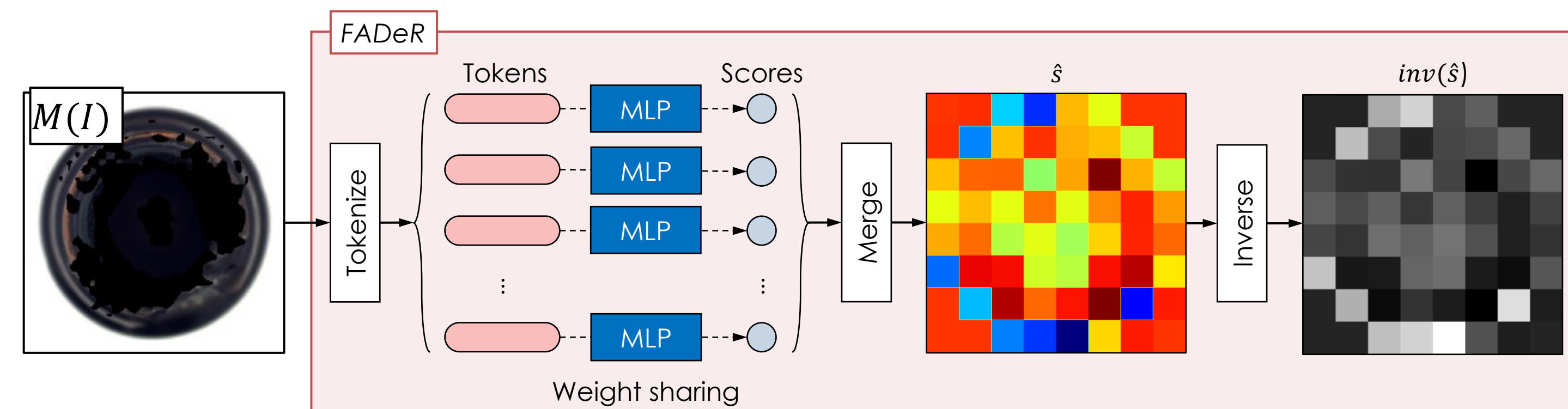
- Challenge to **completely cover anomalous regions** using existing masking methods without additional training
- The model designed to compensate for mask incompleteness **should enable label-free training and scalability**



w/ pre-trained attention model (e.g., ViT)

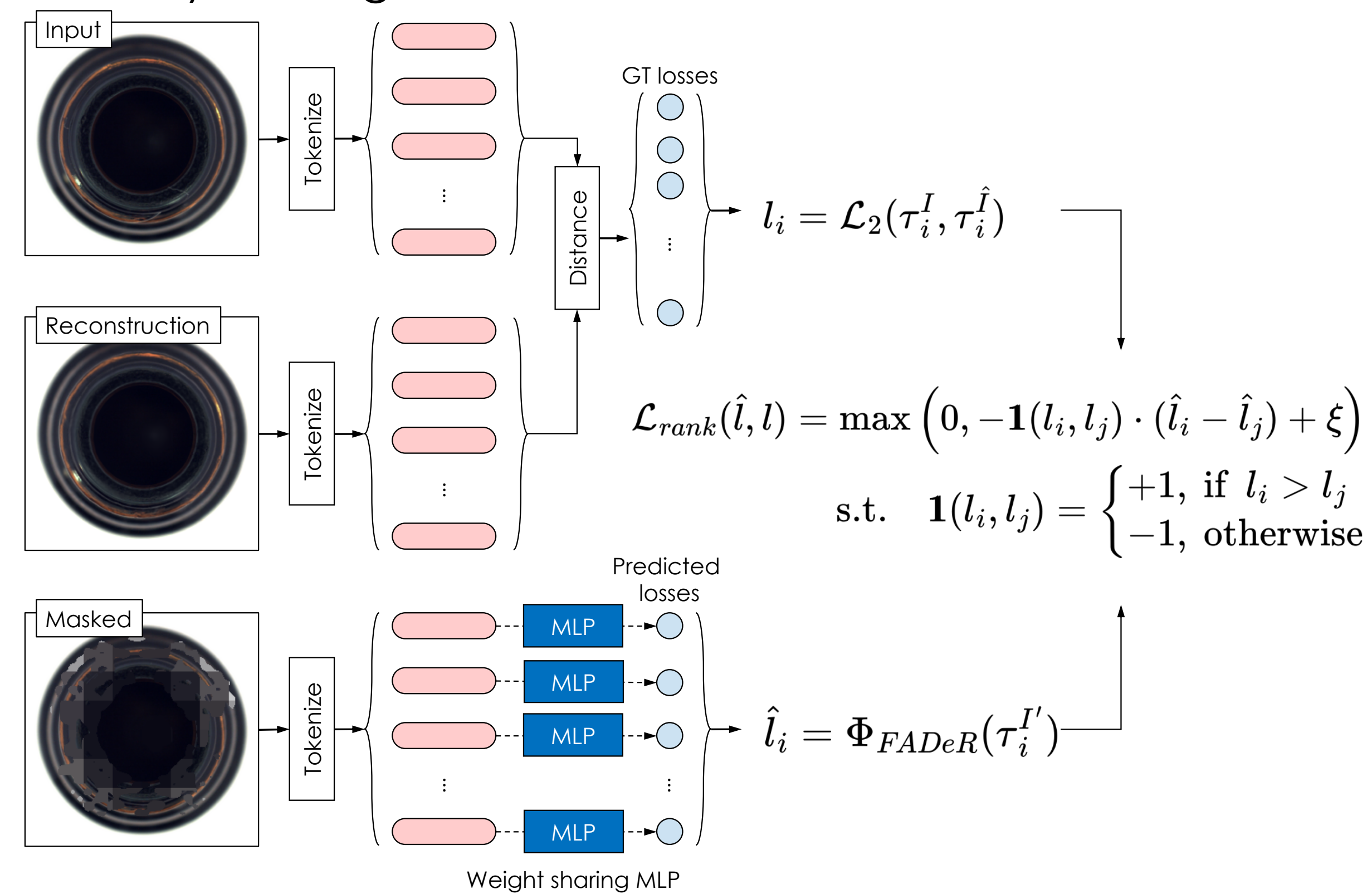
Missing defective parts

- An MLP component designed to overcome incomplete masking by attenuating defective feature representations

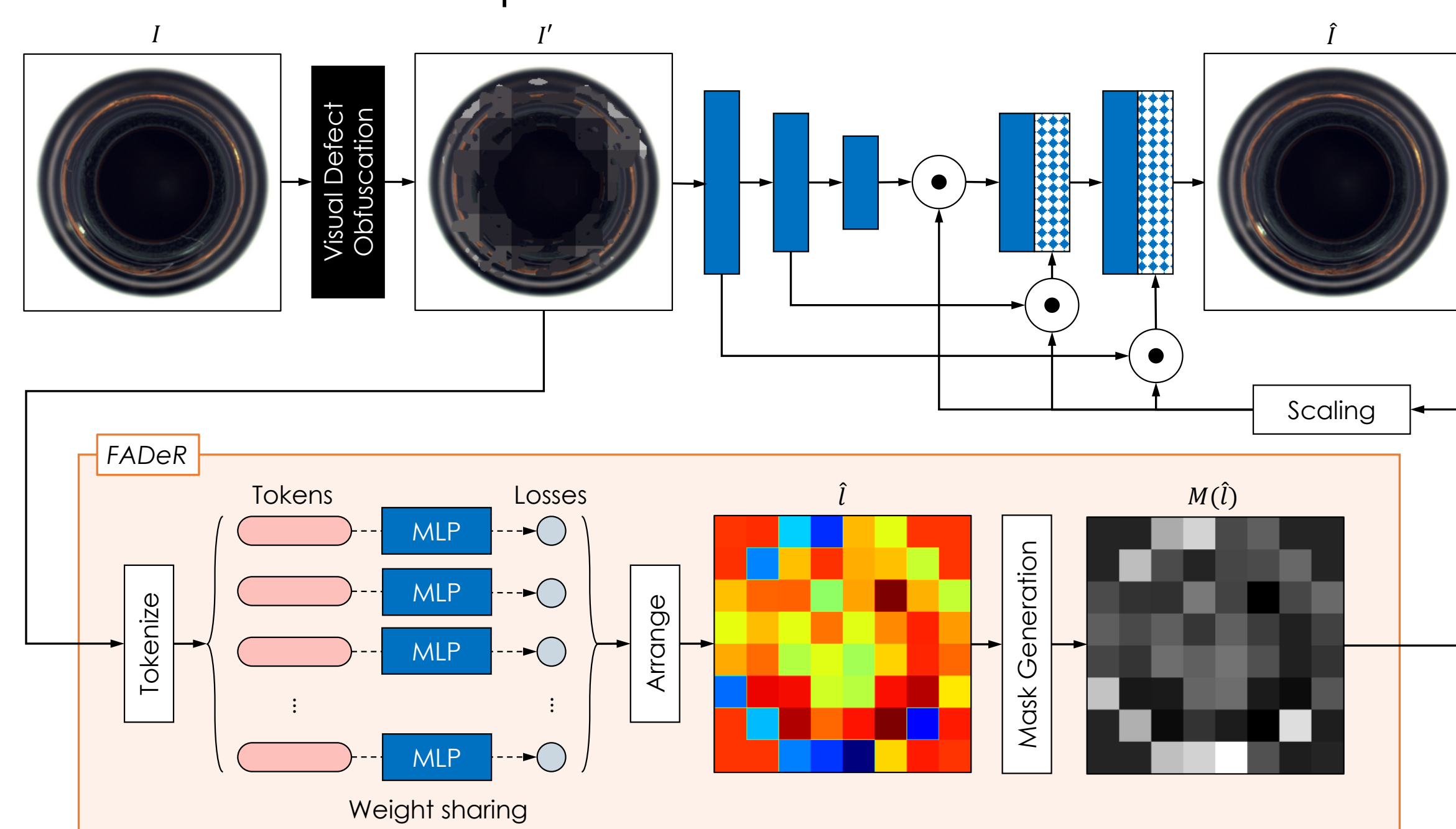


Feature Attenuation of Defective Representation

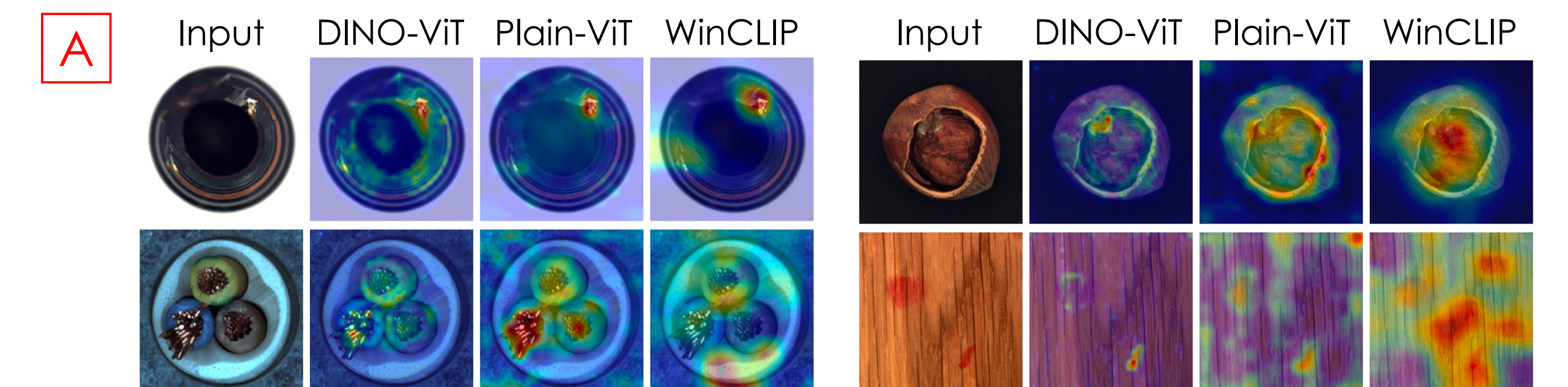
- Enables **label-free training** by leveraging active learning
- FADeR is trained to predict the rank \hat{l} of path-wise reconstruction errors l by ranking loss



- Generates a soft mask **only with a simple two-layered MLP** to attenuate defective representations



Results



B

Model	FADeR _{DINO-EAR}	FADeR _{ViT-EAR}	FADeR _{WinCLIP-EAR}
Bottle	0.997 / 0.914 → 0.998 / 0.951	0.983 / 0.930 → 0.993 / 0.950	0.992 / 0.936
Cable	0.870 / 0.775 → 0.887 / 0.856	0.817 / 0.798 → 0.825 / 0.872	0.838 / 0.550 → 0.794 / 0.612
Capsule	0.870 / 0.944 → 0.947 / 0.980	0.863 / 0.965 → 0.933 / 0.977	0.867 / 0.935 → 0.930 / 0.968
Carpet	0.899 / 0.974 → 0.971 / 0.992	0.701 / 0.945 → 0.756 / 0.981	0.679 / 0.909 → 0.746 / 0.955
Grid	0.959 / 0.974 → 0.983 / 0.986	0.765 / 0.919 → 0.805 / 0.901	0.866 / 0.940 → 0.861 / 0.939
Hazelnut	0.997 / 0.957 → 0.988 / 0.976	0.954 / 0.952 → 0.973 / 0.977	0.946 / 0.947 → 0.959 / 0.976
Leather	1.000 / 0.992 → 1.000 / 0.996	1.000 / 0.981 → 1.000 / 0.993	1.000 / 0.971 → 1.000 / 0.990
Metal nut	0.876 / 0.793 → 0.876 / 0.828	0.880 / 0.767 → 0.903 / 0.804	0.775 / 0.780 → 0.821 / 0.779
Pill	0.922 / 0.875 → 0.976 / 0.945	0.884 / 0.914 → 0.957 / 0.967	0.879 / 0.898 → 0.956 / 0.951
Screw	0.886 / 0.975 → 0.918 / 0.991	0.883 / 0.977 → 0.905 / 0.993	0.862 / 0.953 → 0.897 / 0.968
Tile	0.962 / 0.857 → 1.000 / 0.950	0.961 / 0.853 → 1.000 / 0.940	0.960 / 0.816 → 0.986 / 0.921
Toothbrush	1.000 / 0.953 → 1.000 / 0.987	1.000 / 0.953 → 0.994 / 0.986	1.000 / 0.959 → 1.000 / 0.986
Transistor	0.947 / 0.745 → 0.933 / 0.825	0.891 / 0.679 → 0.908 / 0.740	0.788 / 0.699 → 0.835 / 0.642
Wood	0.985 / 0.875 → 0.996 / 0.818	0.931 / 0.818 → 0.964 / 0.730	0.997 / 0.691 → 0.997 / 0.751
Zipper	0.955 / 0.930 → 0.987 / 0.988	0.932 / 0.909 → 0.938 / 0.951	0.924 / 0.895 → 0.975 / 0.967
Average	0.942 / 0.902 → 0.964 / 0.938	0.896 / 0.891 → 0.924 / 0.918	0.892 / 0.859 → 0.917 / 0.889

D

Model	2-stage			1-stage
	DRAEM	JNLD	OmniAL	FADeR _{EAR} (ours)
Candle	0.823	0.891	0.851	0.927
Capsules	0.773	0.891	0.879	0.919
Cashew	0.942	0.960	0.971	0.967
Chewing gum	0.934	0.985	0.949	0.926
Fryum	1.000	0.932	0.970	0.968
Macaroni1	0.703	0.943	0.969	0.991
Macaroni2	0.713	0.865	0.899	0.965
PCB1	0.713	0.820	0.966	0.967
PCB2	0.897	0.963	0.994	0.987
PCB3	0.731	0.969	0.969	0.974
PCB4	0.913	0.948	0.974	0.999
Pipe fryum	0.941	0.960	0.914	0.933
Average	0.841	0.930	0.942	0.960

C

Model	FADeR _{DSR-EAR}	FADeR _{MSFlow-EAR}	FADeR _{AMI-EAR}
Bottle	0.991 / 0.953 → 0.984 / 0.938	0.971 / 0.931 → 0.973 / 0.951	0.886 / 0.910 → 0.987 / 0.978
Cable	0.891 / 0.908 → 0.834 / 0.947	0.920 / 0.898 → 0.939 / 0.943	0.879 / 0.794 → 0.856 / 0.928
Capsule	0.834 / 0.943 → 0.927 / 0.976	0.861 / 0.957 → 0.938 / 0.976	0.911 / 0.773 → 0.902 / 0.970
Carpet	0.806 / 0.982 → 0.935 / 0.993	0.983 / 0.970 → 0.984 / 0.982	0.983 / 0.932 → 0.785 / 0.903
Grid	0.930 / 0.984 → 0.984 / 0.989	0.971 / 0.971 → 0.960 / 0.984	0.934 / 0.809 → 0.908 / 0.847
Hazelnut	0.989 / 0.954 → 0.975 / 0.964	0.903 / 0.946 → 0.901 / 0.963	0.729 / 0.958 → 0.895 / 0.950
Leather	1.000 / 0.989 → 1.000 / 0.996	1.000 / 0.978 → 1.000 / 0.993	0.816 / 0.823 → 0.998 / 0.956
Metal nut	0.817 / 0.804 → 0.851 / 0.893	0.765 / 0.915 → 0.767 / 0.939	0.918 / 0.545 → 0.933 / 0.835
Pill	0.824 / 0.917 → 0.977 / 0.957	0.855 / 0.924 → 0.974 / 0.963	0.995 / 0.938 → 0.903 / 0.936
Screw	0.752 / 0.987 → 0.751 / 0.993	0.837 / 0.975 → 0.896 / 0.993	0.883 / 0.927 → 0.786 / 0.920
Tile	0.942 / 0.878 → 0.999 / 0.977	0.937 / 0.875 → 0.999 / 0.980	0.934 / 0.840 → 0.948 / 0.867
Toothbrush	0.992 / 0.979 → 1.000 / 0.991	0.964 / 0.964 → 1.000 / 0.986	0.770 / 0.883 → 0.919 / 0.957
Transistor	0.857 / 0.685 → 0.840 / 0.779	0.876 / 0.764 → 0.891 / 0.839	0.821 / 0.950 → 0.915 / 0.884
Wood	0.943 / 0.917 → 0.977 / 0.817	0.955 / 0.900 → 0.943 / 0.904	0.875 / 0.856 → 0.887 / 0.715
Zipper	0.939 / 0.965 → 0.993 / 0.989	0.947 / 0.929 → 0.947 / 0.982	0.970 / 0.967 → 0.902 / 0.934
Average	0.900 / 0.923 → 0.935 / 0.946	0.916 / 0.926 → 0.941 / 0.959	0.887 / 0.860 → 0.902 / 0.905

- A. Attention maps of saliency/defective regions (w/o FADeR)
- B. Pre-trained attention-based masking methods on the MVtec AD
- C. Latent masking methods on the MVtec AD
- D. Comparison with two-stage model on the VisA
- E. Original binary mask and soft mask for feature attenuation

