

Parametric Dimension Reduction by Preserving Local Structure – Supplemental Material

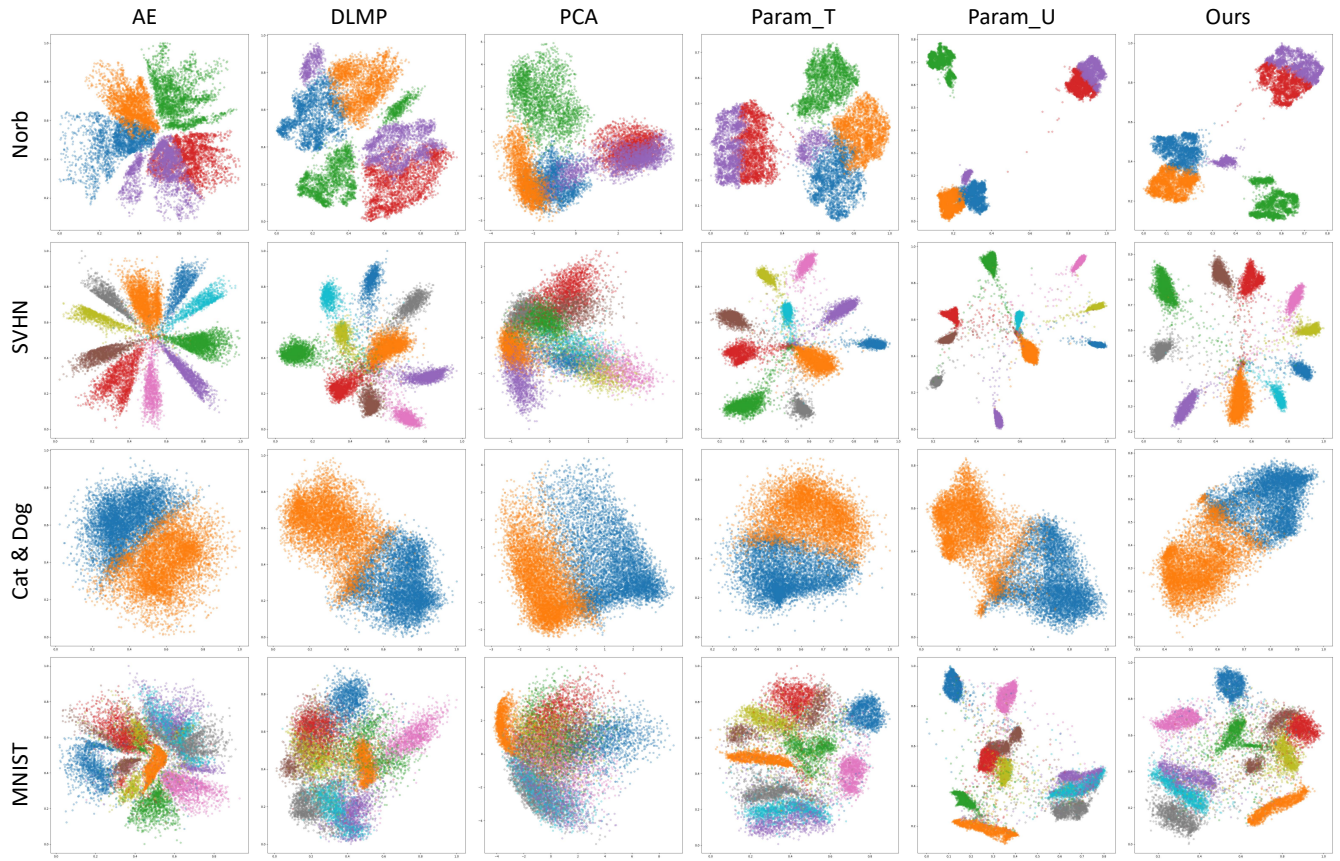


Figure 3: We project high dimensional data to planar space for visual comparison. Each dot is a sample and each distinct color indicates a class.

Width Comparison												
Architecture	Bank	Cat & Dog	Cifar10	Cifar100	F_MNIST	Hatespeech	IMDB	Letter	MNIST	Norb	SVHN	Mean
D \rightarrow 2048 \rightarrow 1024 \rightarrow 512 \rightarrow 256 \rightarrow 2	0.933	0.894	0.966	0.688	0.883	0.866	0.632	0.851	0.906	0.969	0.974	0.869
D \rightarrow 1024 \rightarrow 512 \rightarrow 256 \rightarrow 128 \rightarrow 2	0.933	0.895	0.964	0.683	0.882	0.864	0.633	0.844	0.909	0.967	0.974	0.868
D \rightarrow 512 \rightarrow 256 \rightarrow 128 \rightarrow 64 \rightarrow 2	0.934	0.894	0.962	0.677	0.881	0.861	0.633	0.82	0.895	0.967	0.973	0.863
D \rightarrow 256 \rightarrow 128 \rightarrow 64 \rightarrow 32 \rightarrow 2	0.934	0.894	0.957	0.667	0.879	0.86	0.633	0.795	0.881	0.964	0.97	0.858
Depth Comparison												
Architecture	Bank	Cat & Dog	Cifar10	Cifar100	F_MNIST	Hatespeech	IMDB	Letter	MNIST	Norb	SVHN	Mean
D \rightarrow 1024 \rightarrow 512 \rightarrow 512 \rightarrow 256 \rightarrow 256 \rightarrow 128 \rightarrow 2	0.933	0.894	0.965	0.689	0.882	0.862	0.642	0.845	0.91	0.969	0.973	0.869
D \rightarrow 1024 \rightarrow 512 \rightarrow 256 \rightarrow 128 \rightarrow 2	0.933	0.894	0.963	0.685	0.882	0.864	0.632	0.84	0.904	0.97	0.974	0.867
D \rightarrow 1024 \rightarrow 512 \rightarrow 128 \rightarrow 2	0.933	0.894	0.961	0.677	0.882	0.867	0.633	0.839	0.894	0.968	0.974	0.866
D \rightarrow 1024 \rightarrow 512 \rightarrow 2	0.933	0.893	0.954	0.66	0.88	0.863	0.63	0.808	0.882	0.965	0.971	0.858

Table 2: We compared the dimension reduction results achieved using different network architectures. The numbers are determined using Equation ?? in our main paper. We use background colors, where green to white show the low values to high values, to help readers interpret the results. As can be seen, deeper and wider networks can slightly achieve better results, although with the higher computational costs.

By dataset	Continuity	Method	Bank	Cat & Dog	Cifar10	Cifar100	F_MNIST	Hatespeech	IMDB	Letter	MNIST	Norb	SVHN	Mean
		AE	0.941	0.920	0.970	0.899	0.982	0.881	0.532	0.977	0.948	0.989	0.984	0.911
		DLMP	0.955	0.919	0.964	0.886	0.981	0.926	0.669	0.986	0.953	0.990	0.974	0.928
		PCA	0.947	0.914	0.944	0.864	0.979	0.850	0.414	0.974	0.942	0.973	0.955	0.887
		Param_T	0.959	0.933	0.977	0.923	0.988	0.935	0.769	0.987	0.971	0.992	0.983	0.947
		Param_U	0.966	0.925	0.976	0.919	0.988	0.926	0.733	0.989	0.969	0.991	0.983	0.942
		Ours	0.963	0.932	0.975	0.923	0.987	0.936	0.722	0.989	0.970	0.993	0.983	0.943
	t-SNE	0.968	0.931	0.979	0.936	0.988	0.959	0.814	0.995	0.980	0.997	0.986	0.958	
	Trustworthiness	AE	0.982	0.824	0.957	0.855	0.966	0.897	0.617	0.951	0.913	0.961	0.972	0.900
		DLMP	0.892	0.804	0.925	0.748	0.946	0.757	0.553	0.931	0.819	0.951	0.950	0.843
		PCA	0.900	0.798	0.849	0.683	0.913	0.815	0.607	0.822	0.734	0.873	0.810	0.800
		Param_T	0.967	0.816	0.946	0.830	0.962	0.856	0.600	0.946	0.905	0.942	0.970	0.885
		Param_U	0.941	0.821	0.950	0.838	0.964	0.857	0.596	0.960	0.916	0.950	0.969	0.887
		Ours	0.980	0.825	0.953	0.846	0.965	0.861	0.605	0.954	0.913	0.957	0.970	0.894
		t-SNE	0.995	0.951	0.990	0.974	0.992	0.989	0.807	0.999	0.992	0.999	0.991	0.971
	Neighborhood hit	AE	0.862	0.932	0.925	0.332	0.678	0.810	0.605	0.470	0.717	0.907	0.968	0.746
		DLMP	0.857	0.934	0.862	0.233	0.664	0.786	0.609	0.519	0.711	0.939	0.934	0.732
		PCA	0.849	0.925	0.452	0.176	0.530	0.804	0.596	0.283	0.470	0.784	0.635	0.591
		Param_T	0.861	0.936	0.906	0.290	0.694	0.796	0.617	0.582	0.788	0.941	0.963	0.761
		Param_U	0.860	0.934	0.959	0.318	0.693	0.801	0.630	0.606	0.835	0.951	0.970	0.778
		Ours	0.865	0.937	0.962	0.327	0.701	0.809	0.595	0.645	0.854	0.967	0.970	0.785
t-SNE		0.863	0.939	0.976	0.631	0.803	0.819	0.649	0.925	0.940	0.999	0.980	0.866	
By method	Continuity	AE	0.941	0.920	0.970	0.899	0.982	0.881	0.532	0.977	0.948	0.989	0.984	0.911
		DLMP	0.954	0.907	0.962	0.868	0.973	0.926	0.658	0.986	0.950	0.990	0.971	0.922
		PCA	0.947	0.914	0.944	0.864	0.979	0.850	0.414	0.974	0.942	0.973	0.955	0.887
		Param_T	0.959	0.933	0.977	0.922	0.987	0.935	0.769	0.987	0.971	0.992	0.982	0.947
		Param_U	0.958	0.925	0.976	0.919	0.988	0.924	0.733	0.989	0.969	0.991	0.982	0.941
		Ours	0.963	0.932	0.975	0.923	0.987	0.936	0.722	0.987	0.967	0.993	0.981	0.942
		t-SNE	0.967	0.931	0.976	0.934	0.986	0.959	0.796	0.995	0.979	0.997	0.984	0.955
	Trustworthiness	AE	0.982	0.824	0.957	0.855	0.966	0.897	0.617	0.951	0.913	0.961	0.972	0.900
		DLMP	0.878	0.781	0.919	0.730	0.946	0.756	0.553	0.931	0.819	0.951	0.950	0.838
		PCA	0.900	0.798	0.849	0.683	0.913	0.815	0.607	0.822	0.734	0.873	0.810	0.800
		Param_T	0.967	0.815	0.944	0.814	0.961	0.856	0.600	0.941	0.905	0.942	0.969	0.883
		Param_U	0.941	0.821	0.950	0.838	0.963	0.857	0.594	0.960	0.916	0.946	0.969	0.887
		Ours	0.976	0.825	0.953	0.846	0.965	0.849	0.599	0.949	0.913	0.957	0.970	0.891
		t-SNE	0.995	0.951	0.990	0.974	0.992	0.989	0.807	0.999	0.992	0.999	0.991	0.971
	Neighborhood hit	AE	0.862	0.932	0.925	0.332	0.678	0.810	0.605	0.470	0.717	0.907	0.968	0.746
		DLMP	0.853	0.930	0.857	0.229	0.664	0.778	0.606	0.519	0.651	0.913	0.934	0.721
		PCA	0.849	0.925	0.452	0.176	0.530	0.804	0.596	0.283	0.470	0.784	0.635	0.591
		Param_T	0.851	0.933	0.888	0.290	0.686	0.796	0.602	0.582	0.754	0.894	0.963	0.749
		Param_U	0.860	0.934	0.952	0.314	0.693	0.801	0.630	0.600	0.813	0.935	0.970	0.773
		Ours	0.862	0.937	0.962	0.313	0.701	0.800	0.595	0.645	0.816	0.967	0.968	0.779
		t-SNE	0.863	0.935	0.969	0.624	0.803	0.819	0.642	0.925	0.940	0.999	0.978	0.863

Table 3: We compared our method with baselines on a variety of datasets in terms of continuity, trustworthiness, and neighborhood hit. Similarly, we use background colors, where green to white show the low values to the high values, to help readers interpret the results.