Pyramid-based Scatterplots Sampling for Progressive and Streaming Data Visualization

- Supplementary Material -

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This supplemental material file provides three additional experimental results for our paper: (i) parameter selection, (ii) the case study of the Stock Price data, and (iii) the complete details about the comparative evaluation.



Fig. 1: Parameter analysis on the *Person Activity* dataset [4] with the PDDr and ESRr scores. (a) The opaque scatterplot of the whole 98K data points and (b) rendered as a transparent density map, where some major features are highlighted. The orange dashed boxes and red dashed circles reveal typical low- and medium-density regions, respectively. (c-h) Different sampled results produced by different λ and ω parameters with same *stopLevel*, where each result has around 1.6K data points; the associated specific parameter values and PDDr and ESRr scores are provided under each result. (i-l) Different sampled results produced by increasing *stopLevel* parameter, where the number of data points is decreasing and the associated PDDr and ESRr scores are provided.

1 PARAMETER SELECTION

Fig. 1 shows the influences of parameter settings in terms of the PDDr and ESRr scores on the *Person Activity* dataset [4], where the original scatterplot and density map are shown in Figs. 1(a,b).

Robustness: From Figs. 1(c-l), we can see that the PDDr scores of all results are around 0.9 except Fig. 1(i), indicating that our method is robust in the preservation of relative densities over the parameter choices. This observation is also consistent with the patterns shown in Fig. 1(b), where the densities of the central high-density regions are much larger than those in the surrounding low-density regions.

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Fig. 2: Comparison of our method and the competing methods on the *Person Activity* dataset [4] with the PDDr and ESRr scores. While our method is competitive with these state-of-the-art methods in preserving relative densities, the red dashed boxes shows that our method can visually maintain the structure in the low-density regions more effectively.

 $\lambda \& \omega$: In contrast, the ESRr score decreases as λ increases (see Figs. 1(c,d,e)) or as ω increases (see Figs. 1(f,d,g)), where more low-density regions are preserved in the corresponding results. In our experiment, we empirically choose $\lambda = 0.1$ and $\omega = 0.2$, because they lead to the largest PDDr score and a reasonable ESRr value.

stopLevel: Figs. 1(i,j,k,l) show the influence of the *stopLevel* parameter, where a small *stopLevel* value yields small ESRr and PDDr scores except the PDDr score in Fig. 1(l). This is reasonable since a small *stopLevel* value leads to more low-density regions to be preserved and thus the ESRr and PDDr become small. However, computing the PDDr score further depends on the size of the region to be compared. In this example, the region size 40×40 pixels is close to the sub-region size of the *stopLevel* 6, resulting that the PDDr value at *stopLevel* 6 becomes the largest.

Competing methods: Fig. 2 compares the results generated by our method and the competing methods on this dataset. We can see that the PDDr score of our method is very close to those of random sampling (RS), KD-tree-based sampling (KBS) [2], density-based sampling (DBS) [7], blue noise sampling (BNS) [3], and multi-view z-order sampling (MVZS) [5]. Outlier biased density-based sampling (OBDBS) [8] is slightly worse than our method, while non-uniform sampling (NUS) [1] is the worst. In contrast, NUS performs the best in ESRr, while our method is ranked the second and its score is smaller than all the rest methods. This observation is consistent with the finding revealed by our comparative evaluation presented in the paper.

The results indicate that our pyramid-based sampling is robust for preserving relative densities. Thus, we suggest users to preserve outliers by adjusting λ , ω and *stopLevel* for balancing the data and visibility densities. Our experiments show that setting λ and ω to be around 0.1 and 0.2 is good for most data. A small *stopLevel* may help to show more outliers by adding more samples to low density regions, whereas the proper level can be adjusted for different data sets and user preferences.

2 CASE STUDY: STOCK PRICE

Next, we present two additional density maps of the Stock Price case study in Fig. 3, which is about how the Sep. 11 attacks affected the stock market on the relationship between the stock volume and stock percentage change. We compared two scatterplots (see Figs. 3(a,b)) generated from the dataset for two different time-ranges: (i) from Aug. 11, 2001 to Sep. 10, 2001 (see Fig. 3(a)) and (ii) the whole Sep. 2001 (see Fig. 3(b)). As we can see, the changes in stock percentage have much larger variations in Fig. 3(b) than those in Fig. 3(a). Also, more points with negative stock changes can be observed than those with positive changes. Based on these observations, we see that the Sep. 11 attacks resulted in a negative effect and many stocks show large fluctuations. However, the streaming sampling results in Figs. 3(c,d) show that the horizontal major trend keeps stable around 0% in both ranges, as shown in Figs. 3(e,f). Further comparing the samples in the red and orange boxes reveals more stocks with negative stock changes and large fluctuations in Sep. 2001. Therefore, we conclude that our streaming sampling method produces faithful visualizations.

3 COMPARATIVE EVALUATION

Finally, we elaborate on the comparative evaluation of our static sampling with 12 synthetic datasets created by mixing Gaussian distributions and a uniform distribution, and 28 real datasets collected from Kaggle [6] and the UCI data repository [4]. Specifically, we provide screenshots of the original scatterplot and sampled results generated by our approach (PBS) and seven competitive sampling methods: random sampling (RS), blue noise sampling (BNS) [3], density-based sampling (DBS) [7], non-uniform sampling (NUS) [1], outlier biased density-based sampling (OBDBS) [8], multi-view z-order sampling (MVZS) [5], and KD-tree-based sampling (KBS) [2] with their corresponding scores.



Fig. 3: Scatterplots that show the relationship between stock volume (horizontal) and stock percentage change (vertical) of the historical stock market dataset for two different time ranges: before the September 11 attacks (left column) and the whole September 2001 (right column). (a,b) the overplotted scatterplots of the original data; (c,d) streaming visualization results of our method from (a,b); and (e,f) the corresponding density maps of (a,b).

			aba	one				
Input	PE	S	NU	JS	R	S	KE	3S
	1		P		1		P	an takat Arita Arita an
# points: 4177	PDDr: 0.93	ESRr: 0.04	PDDr: 0.91	ESRr: 0.00	PDDr: 0.93	ESRr: 0.06	PDDr: 0.94	ESRr: 0.03
Density map	BN	IS	DE	35	OBI	OBS	M٧	zs
	1		1					
# samples: ~1091	PDDr: 0.94	ESRr: 0.05	PDDr: 0.93	ESRr: 0.07	PDDr: 0.91	ESRr: 0.05	PDDr: 0.95	ESRr: 0.06
			av	ila				
Input	PE	S	NU	JS	R	S	KE	3S
# points: 5672	PDDr: 0.94	ESRr: 0.08	PDDr: 0.88	ESRr: 0.00	PDDr: 0.94	ESRr: 0.14	PDDr: 0.93	ESRr: 0.10
Density map	BN	15	DE	3S	OBI	OBS	M٧	ZS
# samples: ~1366	PDDr: 0.94	ESRr: 0.11	PDDr: 0.93	ESRr: 0.13	PDDr: 0.91	ESRr: 0.09	PDDr: 0.94	ESRr: 0.13
			Barce	elona				
Input	PE	S	NU	JS	R	S	KE	S
# points: 10284	PDDr: 0.84	ESRr: 0.06	PDDr: 0.67	ESRr: 0.01	PDDr: 0.83	ESRr: 0.11	PDDr: 0.86	ESRr: 0.06
Density map	BN	15	DE	35	OBI	OBS	M٧	ZS
-	-							



			DailyS	ports				
Input	PB	S	NU	JS	R	S	KB	S
# points: 540000	PDDr: 0.81	ESRr: 0.13	PDDr: 0.70	ESRr: 0.05	PDDr: 0.91	ESRr: 0.22	PDDr: 0.87	ESRr: 0.17
Density map	BN	IS	DE	3S	OBE	DBS	MV	ZS
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# samples: ~1890	PDDr: 0.75	ESRr: 0.15	PDDr: 0.90	ESRr: 0.21	PDDr: 0.89	ESRr: 0.22	PDDr: 0.91	ESRr: 0.21
			diab	etes				
Input	PB	S	NU	JS	R	S	KB	S
# points: 99493	PDDr: 0.87	ESRr: 0.26	PDDr: 0.46	ESRr: 0.05	PDDr: 0.87	ESRr: 0.33	PDDr: 0.90	ESRr: 0.31
Density map	BN	IS	DB	S	OBE	DBS	MV	ZS
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C BERRING AND								
# samples: ~844	PDDr: 0.84	ESRr: 0.23	PDDr: 0.86	ESRr: 0.34	PDDr: 0.83	ESRr: 0.33	PDDr: 0.89	ESRr: 0.33
# samples: ~844	PDDr: 0.84	ESRr: 0.23	PDDr: 0.86 Emergency	ESRr: 0.34 /_911Calls	PDDr: 0.83	ESRr: 0.33	PDDr: 0.89	ESRr: 0.33
# samples: ~844	PDDr: 0.84	ESRr: 0.23	PDDr: 0.86 Emergency	ESRr: 0.34 /_ 911Calls JS	PDDr: 0.83	ESRr: 0.33 S	PDDr: 0.89	ESRr: 0.33
# samples: ~844	PDDr: 0.84	ESRr: 0.23	PDDr: 0.86 Emergency	ESRr: 0.34 /_911Calls JS	PDDr: 0.83	ESRr: 0.33 S	PDDr: 0.89 KB	ESRr: 0.33
# samples: ~844 Input # points: 332591	PDDr: 0.84 PE	ESRr: 0.23 S ESR: 0.23 ESR: 0.07	PDDr: 0.86 Emergency NL	ESRr: 0.34 2911Calls JS ESRr: 0.00	PDDr: 0.83 R PDDr: 0.90	ESRr: 0.33 S ESRr: 0.11	PDDr: 0.89 KB	ESRr: 0.33
# samples: ~844 Input # points: 332591 Density map	PDDr: 0.84 PE	ESRr: 0.23 S S ESRr: 0.07 IS	PDDr: 0.86 Emergency NU PDDr: 0.74	ESRr: 0.34 /_911Calls JS ESRr: 0.00 SS	PDDr: 0.83 R PDDr: 0.90 PDDr: 0.90	ESRr: 0.33 S ESRr: 0.11 DBS	PDDr: 0.89 KB	ESRr: 0.33
# samples: ~844 Input # points: 332591 Density map	PDDr: 0.84 PE	ESRr: 0.23 S ESRr: 0.07 IS	PDDr: 0.86 Emergency NU PDDr: 0.74 PDDr: 0.74	ESRr: 0.34 /_911Calls JS ESRr: 0.00 BS	PDDr: 0.83 R PDDr: 0.90 OBI	ESRr: 0.33 S ESRr: 0.11 DBS	PDDr: 0.89 KB	ESRr: 0.33

			Epile	eptic				
Input	PE	3S	N	US	R	S	KE	3S
					200			
# points: 11500	PDDr: 0.88	ESRr: 0.01	PDDr: 0.88	ESRr: 0.01	PDDr: 0.86	ESRr: 0.03	PDDr: 0.90	ESRr: 0.01
Density map	BN	1S	DI	BS	OBI	OBS	M۷	ZS
# samples: ~4824	PDDr: 0.87	ESRr: 0.01	PDDr: 0.85	ESRr: 0.02	PDDr: 0.75	ESRr: 0.02	PDDr: 0.90	ESRr: 0.01
			fac	ial				
Input	PE	S	N	US	R	S	KE	3S
# points: 12903	PDDr: 0.91	ESRr: 0.08	PDDr: 0.82	ESRr: 0.01	PDDr: 0.91	ESRr: 0.16	PDDr: 0.91	ESRr: 0.12
Density map	BN	1S	DI	BS	OBI	OBS	M٧	'ZS
# samples: ~1448	PDDr: 0.92	ESRr: 0.10	PDDr: 0.91	ESRr: 0.14	PDDr: 0.91	ESRr: 0.15	PDDr: 0.91	ESRr: 0.14
			forest_co	overtype				
Input	PE	S	N	US	R	S	KE	S
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# points: 41802	PDDr: 0.92	ESRr: 0.08	PDDr: 0.84	ESRr: 0.01	PDDr: 0.93	ESRr: 0.18	PDDr: 0.93	ESRr: 0.14
Density map	BN	15	DI	BS	OBI	OBS	M٧	ZS
# samples: ~2534	PDDr: 0.91	ESRr: 0.10	PDDr: 0.93	ESRr: 0.18	PDDr: 0.93	ESRr: 0.16	PDDr: 0.95	ESRr: 0.17

			gest	ture				
Input	PB	S	NU	JS	R	S	KE	3S
# points: 4845	PDDr: 0.87	ESRr: 0.07	PDDr: 0.78	ESRr: 0.01	PDDr: 0.88	ESRr: 0.13	PDDr: 0.87	ESRr: 0.08
Density map	BN	IS	DI	3S	OBI	OBS	M٧	'ZS
# samples: ~1038	PDDr: 0.87	ESRr: 0.08	PDDr: 0.88	ESRr: 0.12	PDDr: 0.85	ESRr: 0.13	PDDr: 0.90	ESRr: 0.13
			google_s	entiment				
Input	PB	S	NUS		R	S	KE	3S
# points: 37427	PDDr: 0.82	ESRr: 0.16	PDDr: 0.72	ESRr: 0.03	PDDr: 0.82	ESRr: 0.28	PDDr: 0.83	ESRr: 0.15
Density map	BN	IS	DI	35	OBI	OBS	M۷	ZS
# samples: ~2165	PDDr: 0.82	ESRr: 0.15	PDDr: 0.83	ESRr: 0.27	PDDr: 0.79	ESRr: 0.24	PDDr: 0.84	ESRr: 0.23
			HT_se	ensor				
Input	PE	S	NU	JS	R	S	KE	3S
# points: 1565007	PDDr: 0.90	ESRr: 0.11	PDDr: 0.81	ESRr: 0.04	PDDr: 0.91	ESRr: 0.18	PDDr: 0.93	ESRr: 0.15
Density map	BN	IS	DI	35	OBI	OBS	MV	'ZS
# samples: ~2320	PDDr: 0.86	ESRr: 0.15	PDDr: 0.88	ESRr: 0.23	PDDr: 0.87	ESRr: 0.21	PDDr: 0.88	ESRr: 0.23

			mr	nist				
Input	PB	S	N	US	R	S	KE	S
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# points: 70000	PDDr: 0.84	ESRr: 0.02	PDDr: 0.87	ESRr: 0.02	PDDr: 0.83	ESRr: 0.03	PDDr: 0.87	ESRr: 0.02
Density map	BN	IS	DI	BS	OBE	OBS	MV	ZS
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# samples: ~9534	PDDr: 0.83	ESRr: 0.02	PDDr: 0.81	ESRr: 0.04	PDDr: 0.70	ESRr: 0.03	PDDr: 0.90	ESRr: 0.03
			Мо	Сар				
Input	PB	S	N	US	R	S	KE	S
# points: 74975	PDDr: 0.91	ESRr: 0.17	PDDr: 0.70	ESRr: 0.03	PDDr: 0.92	ESRr: 0.26	PDDr: 0.93	ESRr: 0.23
Density map	BN	IS	DI	BS	OBI	OBS	MV	ZS
# samples: ~1223	PDDr: 0.91	ESRr: 0.18	PDDr: 0.91	ESRr: 0.25	PDDr: 0.89	ESRr: 0.27	PDDr: 0.91	ESRr: 0.24
			NewsPo	pularity				
Input	PB	S	N	US	R	S	KE	S
					S. C.			
# points: 93239	PDDr: 0.76	ESRr: 0.01	PDDr: 0.80	ESRr: 0.02	PDDr: 0.74	ESRr: 0.02	PDDr: 0.80	ESRr: 0.01
Density map	BN	IS	DI	BS	OBI	OBS	MV	ZS
# samples: ~6279	PDDr: 0.76	ESRr: 0.02	PDDr: 0.73	ESRr: 0.02	PDDr: 0.56	ESRr: 0.02	PDDr: 0.78	ESRr: 0.01

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Input	PE	3S	N	JS	R	S	KE	S
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# points: 34410	PDDr: 0.77	ESRr: 0.06	PDDr: 0.63	ESRr: 0.03	PDDr: 0.83	ESRr: 0.11	PDDr: 0.83	ESRr: 0.04
Density map	BN	1S	DI	BS	OBI	OBS	Mν	ZS
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# samples: ~722	PDDr: 0.80	ESRr: 0.06	PDDr: 0.80	ESRr: 0.11	PDDr: 0.80	ESRr: 0.11	PDDr: 0.82	ESRr: 0.10
			PersonActiv	ity_4classes				
Input	PE	S	NUS		R	S	KE	S
						250.0		
# points: 98568	PDDr: 0.94	ESRr: 0.21	PDDr: 0.83	ESRr: 0.02	PDDr: 0.94	ESRr: 0.33	PDDr: 0.96	ESRr: 0.29
Density map	BN	15	DI	BS	OBI	OBS	Mν	ZS
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# samples: ~2878	PDDr: 0.93	ESRr: 0.24	PDDr: 0.95	ESRr: 0.31	PDDr: 0.93	ESRr: 0.32	PDDr: 0.95	ESRr: 0.32
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Input	PE	S	N	JS	R	S	KE	S
# points: 4435	PDDr: 0.92	ESRr: 0.04	PDDr: 0.89	ESRr: 0.00	PDDr: 0.90	ESRr: 0.08	PDDr: 0.92	ESRr: 0.06
Density map	BN	15	DI	BS	OBI	OBS	MV	ZS
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# samples: ~2033	PDDr: 0.91	ESRr: 0.07	PDDr: 0.91	ESRr: 0.08	PDDr: 0.90	ESRr: 0.09	PDDr: 0.92	ESRr: 0.08

			senso	orless				
Input	PE	S	NU	JS	R	S	KE	3S
# points: 42552	PDDr: 0.92	ESRr: 0.07	PDDr: 0.81	ESRr: 0.01	PDDr: 0.92	ESRr: 0.08	PDDr: 0.94	ESRr: 0.06
Density map	BN	15	DI	BS	OBI	OBS	M۷	ZS
# samples: ~1986	PDDr: 0.92	ESRr: 0.06	PDDr: 0.93	ESRr: 0.08	PDDr: 0.92	ESRr: 0.08	PDDr: 0.95	ESRr: 0.09
			Spo	tify				
Input	PE	S	NU	JS	R	S	KE	3S
# points: 174389	PDDr: 0.93	ESRr: 0.15	PDDr: 0.84	ESRr: 0.02	PDDr: 0.92	ESRr: 0.20	PDDr: 0.95	ESRr: 0.16
Density map	BN	15	DI	BS	OBI	OBS	M٧	'ZS
# samples: ~3481	PDDr: 0.90	ESRr: 0.16	PDDr: 0.91	ESRr: 0.19	PDDr: 0.90	ESRr: 0.18	PDDr: 0.95	ESRr: 0.18
			synth	etic1				
Input	PE	S	NU	JS	R	S	KE	S
# points: 255147	PDDr: 0.94	ESRr: 0.28	PDDr: 0.78	ESRr: 0.03	PDDr: 0.95	ESRr: 0.38	PDDr: 0.96	ESRr: 0.34
Density map	BN	15	DI	BS	OBE	OBS	M۷	'ZS
# samples: ~2170	PDDr: 0.94	ESRr: 0.29	PDDr: 0.95	ESRr: 0.36	PDDr: 0.86	ESRr: 0.37	PDDr: 0.96	ESRr: 0.37

			synthe	etic10				
Input	PE	3S	NU	JS	R	S	KE	3S
# points: 45698	PDDr: 0.94	ESRr: 0.18	PDDr: 0.79	ESRr: 0.00	PDDr: 0.96	ESRr: 0.25	PDDr: 0.95	ESRr: 0.16
Density map	BN	1S	DE	3S	OBE	OBS	M٧	ZS
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# samples: ~1646	PDDr: 0.95	ESRr: 0.18	PDDr: 0.95	ESRr: 0.25	PDDr: 0.89	ESRr: 0.24	PDDr: 0.97	ESRr: 0.24
			synthe	etic11				
Input	PE	3S	NU	JS	R	S	KE	3S
						.		
# points: 51416	PDDr: 0.94	ESRr: 0.18	PDDr: 0.75	ESRr: 0.02	PDDr: 0.93	ESRr: 0.24	PDDr: 0.94	ESRr: 0.18
Density map	BN	15	DE	3S	OBE	OBS	M۷	ZS
•								
# samples: ~1283	PDDr: 0.92	ESRr: 0.17	PDDr: 0.93	ESRr: 0.23	PDDr: 0.86	ESRr: 0.23	PDDr: 0.95	ESRr: 0.23
			synthe	etic12				
Input	PE	S	NU	JS	R	S	KE	S
# points: 2001000	PDDr: 0.96	ESRr: 0.49	PDDr: 0.77	ESRr: 0.09	PDDr: 0.96	ESRr: 0.57	PDDr: 0.97	ESRr: 0.56
Density map	BN	1S	DE	3S	OBE	OBS	M٧	ZS
•••								
# samples: ~2227	PDDr: 0.95	ESRr: 0.58	PDDr: 0.95	ESRr: 0.59	PDDr: 0.91	ESRr: 0.60	PDDr: 0.97	ESRr: 0.56

			synth	etic2				
Input	PE	S	NU	JS	R	S	KE	3S
# points: 10000	PDDr: 0.94	ESRr: 0.05	PDDr: 0.89	ESRr: 0.00	PDDr: 0.93	ESRr: 0.11	PDDr: 0.94	ESRr: 0.05
Density map	BN	15	DI	35	OBI	OBS	M٧	ZS
Constant of Consta								
C. Materia								
# samples: ~2806	PDDr: 0.94	ESRr: 0.08	PDDr: 0.93	ESRr: 0.08	PDDr: 0.89	ESRr: 0.09	PDDr: 0.95	ESRr: 0.10
			synth	etic3				
Input	PE	S	N	JS	R	S	KE	S
# points: 10000	PDDr: 0.91	ESRr: 0.05	PDDr: 0.88	ESRr: 0.00	PDDr: 0.91	ESRr: 0.10	PDDr: 0.94	ESRr: 0.06
Density map	BN	15	DI	35	OBI	OBS	M٧	'ZS
# samples: ~3138	PDDr: 0.92	ESRr: 0.07	PDDr: 0.92	ESRr: 0.08	PDDr: 0.87	ESRr: 0.06	PDDr: 0.94	ESRr: 0.09
			synth	etic4				
Input	PE	S	N	JS	R	S	KE	S
# points: 30000	PDDr: 0.91	ESRr: 0.03	PDDr: 0.88	ESRr: 0.01	PDDr: 0.90	ESRr: 0.06	PDDr: 0.93	ESRr: 0.03
Density map	BN	15	DI	35	OBI	OBS	M٧	ZS
# samples: ~6035	PDDr: 0.90	ESRr: 0.04	PDDr: 0.90	ESRr: 0.05	PDDr: 0.82	ESRr: 0.04	PDDr: 0.93	ESRr: 0.04

			synth	etic5				
Input	PB	S	NU	JS	R	S	KE	3S
# points: 30000	PDDr: 0.93	ESRr: 0.08	PDDr: 0.86	ESRr: 0.00	PDDr: 0.94	ESRr: 0.14	PDDr: 0.95	ESRr: 0.12
Density map	BN	IS	DE	35	OBE	DBS	MV	ZS
800								
# samples: ~3545	PDDr: 0.93	ESRr: 0.11	PDDr: 0.93	ESRr: 0.13	PDDr: 0.88	ESRr: 0.11	PDDr: 0.95	ESRr: 0.15
			synth	etic6				
Input	PB	S	NU	JS	R	S	KE	3S
		e e de la Co		e en de la Re				e en de la Re
# points: 37606	PDDr: 0.96	ESRr: 0.21	PDDr: 0.80	ESRr: 0.02	PDDr: 0.95	ESRr: 0.27	PDDr: 0.93	ESRr: 0.14
Density map	BN	IS	DE	35	OBE	DBS	MV	ZS
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				113.71				111
					a the second sec	А.		
# samples: ~879	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95	ESRr: 0.27	PDDr: 0.87	ESRr: 0.26	PDDr: 0.95	ESRr: 0.26
# samples: ~879	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95	ESRr: 0.27	PDDr: 0.87	ESRr: 0.26	PDDr: 0.95	ESRr: 0.26
# samples: ~879	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 Synth N	ESRr: 0.27 etic7 JS	PDDr: 0.87	ESRr: 0.26 S	PDDr: 0.95	ESRr: 0.26
# samples: ~879	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 Synth NI	ESRr: 0.27 etic7 JS	PDDr: 0.87	ESRr: 0.26 S	PDDr: 0.95	ESRr: 0.26
# samples: ~879	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 Synth NU	ESRr: 0.27 etic7 JS	PDDr: 0.87	ESRr: 0.26	PDDr: 0.95	ESRr: 0.26
# samples: ~879 Input # points: 37177	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 Synth NI PDDr: 0.78	ESRr: 0.27 etic7 JS ESRr: 0.02	PDDr: 0.87 R PDDr: 0.95	ESRr: 0.26 S ESRr: 0.21	PDDr: 0.94	ESRr: 0.26
# samples: ~879 Input # points: 37177 Density map	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 Synth NI PDDr: 0.78 PDDr: 0.78	ESRr: 0.27 etic7 JS ESRr: 0.02	PDDr: 0.87 R PDDr: 0.95 PDDr: 0.95	ESRr: 0.26 S ESRr: 0.21 DBS	PDDr: 0.95	ESRr: 0.26 35 ESRr: 0.17 ZS
# samples: ~879 Input # points: 37177 Density map	PDDr: 0.94	ESRr: 0.20 S ESRr: 0.17 IS	PDDr: 0.95 Synth NI PDDr: 0.78 DE	ESRr: 0.27 etic7 JS ESRr: 0.02 3S	PDDr: 0.87 PDDr: 0.95 OBE	ESRr: 0.26 S ESRr: 0.21 DSS	PDDr: 0.94 PDDr: 0.94	ESRr: 0.26 35 ESRr: 0.17 ZS
# samples: ~879 Input # points: 37177 Density map	PDDr: 0.94	ESRr: 0.20	PDDr: 0.95 synth NI PDDr: 0.78 PDDr: 0.78	ESRr: 0.27 etic7 JS ESRr: 0.02 3S	PDDr: 0.87 R PDDr: 0.95 OBI	ESRr: 0.26 S ESRr: 0.21 DBS	PDDr: 0.95	ESRr: 0.26

			synth	etic8				
Input	PB	S	NU	JS	R	S	KE	3S
# points: 40517	PDDr: 0.95	ESRr: 0.18	PDDr: 0.82	ESRr: 0.02	PDDr: 0.94	ESRr: 0.24	PDDr: 0.94	ESRr: 0.17
Density map	BN	IS	DI	3S	OBI	OBS	M۷	ZS
					*		***	
# samples: ~1194	PDDr: 0.94	ESRr: 0.18	PDDr: 0.94	ESRr: 0.23	PDDr: 0.86	ESRr: 0.24	PDDr: 0.96	ESRr: 0.23
	1		synth	etic9	1		1	
Input	PB	S	NU	JS	R	S	KE	3S
# points: 37423	PDDr: 0.96	ESRr: 0.13	PDDr: 0.81	ESRr: 0.00	PDDr: 0.95	ESRr: 0.19	PDDr: 0.95	ESRr: 0.10
Density map	BN	IS	DI	3S	OBI	OBS	M٧	ZS
. *.					-i.,		- (۰.
					*			
# samples: ~1206	PDDr: 0.95	ESRr: 0.15	PDDr: 0.95	ESRr: 0.19	PDDr: 0.89	ESRr: 0.19	PDDr: 0.97	ESRr: 0.18
			terro	orism				
Input	PB	S	NU	JS	R	S	KE	3S
			and an				an a	
# points: 177133	PDDr: 0.90	ESRr: 0.18	PDDr: 0.62	ESRr: 0.05	PDDr: 0.89	ESRr: 0.26	PDDr: 0.91	ESRr: 0.20
Density map	BN	IS	DI	3S	OBE	OBS	M٧	ZS
					n an			
# samples: ~465	PDDr: 0.89	ESRr: 0.18	PDDr: 0.89	ESRr: 0.25	PDDr: 0.87	ESRr: 0.27	PDDr: 0.90	ESRr: 0.24

			TLC	trip				
Input	PE	S	NU	JS	R	S	KE	3S
					ļ			
# points: 1999697	PDDr: 0.98	ESRr: 0.32	PDDr: 0.61	ESRr: 0.10	PDDr: 0.98	ESRr: 0.49	PDDr: 0.98	ESRr: 0.47
Density map	BN	IS	DE	35	OBI	OBS	M۷	'ZS
	,							
# samples: ~628	PDDr: 0.98	ESRr: 0.49	PDDr: 0.95	ESRr: 0.50	PDDr: 0.97	ESRr: 0.49	PDDr: 0.98	ESRr: 0.49
			uk_road	l_safety				
Input	PB	S	NU	JS	R	S	KE	3S
# points: 2047256	PDDr: 0.94	ESRr 0.02	PDDr: 0.88	ESBr: 0.00		ESDr. 0.04		
		ESTAT: 0.0E	1001.0.00	ESI(1: 0.00	PDDI: 0.90	L3KI. 0.04	PDDF: 0.90	ESRI: 0.04
Density map	BN	IS	DE	35	OBI	DBS	MV	ZS 25
Density map	BN	IS	DI	35	OBI	DBS	MV	25 25
Density map # samples: ~1012	PDDr: 0.95	ESRr: 0.05	PDDr: 0.95	ESRr: 0.04	PDDr: 0.95	ESR: 0.04	PDDr: 0.90 MV PDDr: 0.95	ESRr: 0.04
Density map	BN PDDr: 0.95	ESRr: 0.05	PDDr: 0.95	ESRr: 0.04 ensus	PDDr: 0.95	ESR: 0.04	PDDr: 0.90 MV PDDr: 0.95	ESRr: 0.04
Density map # samples: ~1012 Input	PDDr: 0.95	ESRr: 0.05	PDDr: 0.95 US_cce	ESRr: 0.04 ESRr: 0.04 ESSR: JS	PDDr: 0.95	ESR: 0.04 DBS ESR: 0.04	PDDr: 0.90 MV PDDr: 0.95	ESRr: 0.04 ESRr: 0.05
Density map # samples: ~1012	BN PDDr: 0.95 PE	ESRI: 0.05 ESRI: 0.05	PDDr: 0.95	ESRr: 0.04 ESRr: 0.04 ESS	PDDr: 0.95	ESR: 0.04 DBS ESR: 0.04 S	PDDr: 0.90 MV PDDr: 0.95	ESRr: 0.04 (ZS) (ESRr: 0.05) (SS)
Density map # samples: ~1012 Input # points: 2000000	BN PDDr: 0.95 PE	ESRr: 0.05 ESRr: 0.25	PDDr: 0.95	ESRr: 0.04 ESRr: 0.04 ESRr: 0.04 ESRr: 0.13	PDDr: 0.93	ESR: 0.04 DBS ESR: 0.04 S ESR: 0.04 ESR: 0.04	PDDr: 0.90 MV PDDr: 0.95 KE PDDr: 0.91	ESRr: 0.04 ESRr: 0.05
Density map # samples: ~1012 Input # points: 2000000 Density map	PDDr: 0.95 PDDr: 0.84 BN	ESRr: 0.05 S ESRr: 0.25 IS	PDDr: 0.95 US_cce NU PDDr: 0.74	ESRr: 0.04 ESRr: 0.04 ESRr: 0.04 ESRr: 0.13 ESRr: 0.13	PDDr: 0.95 PDDr: 0.95 PDDr: 0.87 PDDr: 0.87	ESRI: 0.04 DBS ESRI: 0.04 S ESRI: 0.04 ESRI: 0.04	PDDr: 0.90 MV PDDr: 0.95 KE PDDr: 0.91 MV	ESRr: 0.04 (ZS) (ESRr: 0.05) (ESRr: 0.28) (ZS)
Density map # samples: ~1012 Input # points: 2000000 Density map	BN PDDr: 0.95 PDDr: 0.84 PDDr: 0.84	ESRr: 0.05 S ESRr: 0.25 IS	PDDr: 0.95 US_cce NU PDDr: 0.74	ESRr: 0.04 ESRr: 0.04 ESRr: 0.13 ESRr: 0.13	PDDr: 0.95 PDDr: 0.95 PDDr: 0.87 OBI	ESR: 0.04 ESR: 0.04 S ESR: 0.04 S ESR: 0.34 DBS	PDDr: 0.90 MV PDDr: 0.95 PDDr: 0.91 MV	ESRr: 0.04 ZS ESRr: 0.05 3S ESRr: 0.28 ZS

vehicles								
Input	PBS		NUS		RS		KBS	
# points: 383430	PDDr: 0.90	ESRr: 0.17	PDDr: 0.79	ESRr: 0.05	PDDr: 0.89	ESRr: 0.29	PDDr: 0.91	ESRr: 0.22
Density map	BNS		DBS		OBDBS		MVZS	
# samples: ~4027	PDDr: 0.90	ESRr: 0.24	PDDr: 0.89	ESRr: 0.27	PDDr: 0.82	ESRr: 0.26	PDDr: 0.92	ESRr: 0.27

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