

# Declutter and resample: Towards parameter free denoising.

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Joint work with:

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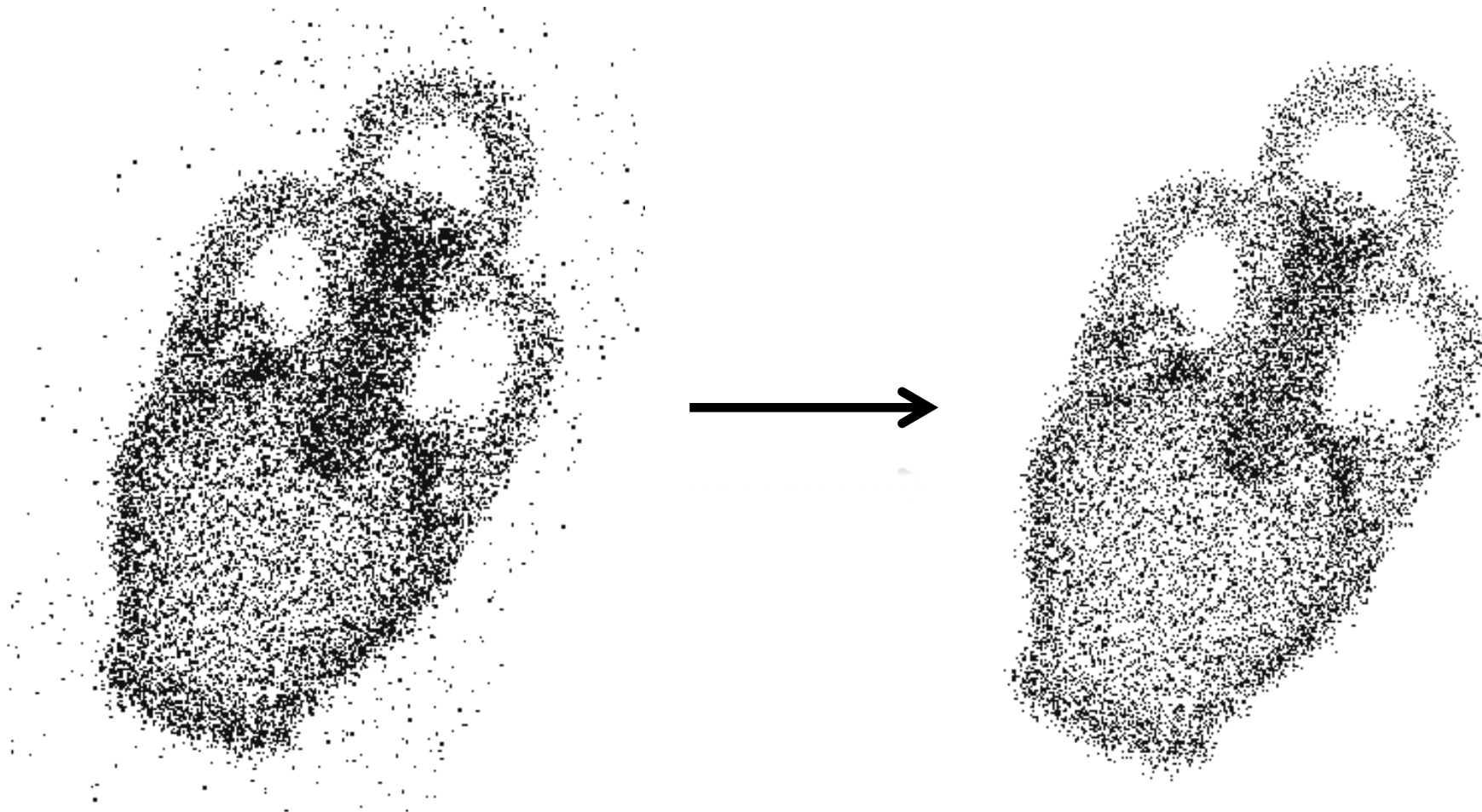
Tamal K. Dey

Yusu Wang

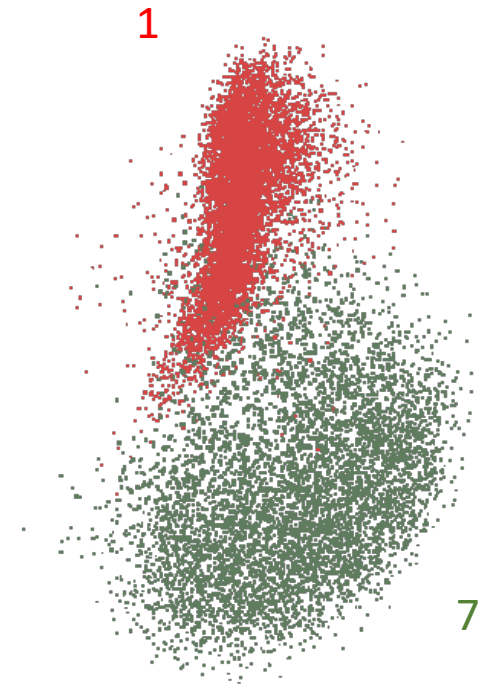
# Outline-Parameter-free denoising algorithm

- Parameter-free denoising algorithm
  - Introduction
  - Preliminaries
  - Declutter algorithm
  - Parameter-free algorithm
  - Discussions

# Introduction



# Introduction



# Introduction

- Deconvolution - noise model/parameter
- Thresholding - parameter

# Outline-Parameter-free denoising algorithm

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  - Introduction
  - Preliminaries
  - Declutter algorithm
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  - Discussions

# Preliminaries – sampling condition

- Metric space  $(\mathbb{X}, d_{\mathbb{X}})$
- $k$ -distance to a point set  $P$  :  $d_{P,k}(x) = \sqrt{\frac{1}{k} \sum_{i=1}^k d_{\mathbb{X}}(x, p_i(x))^2}$
- $P$  is an  $\epsilon_k$ -noisy sample of  $K$  if:
  - $\forall x \in K, d_{P,k}(x) \leq \epsilon_k$
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# Preliminaries – adaptive sampling condition

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- $P$  is an  $\epsilon_k$ -adaptive noisy sample of  $K$  if:
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  - $\forall y \in \mathbb{X}, d_{\mathbb{X}}(y, K) \leq d_{P,k}(y) + \epsilon_k f(\bar{y})$

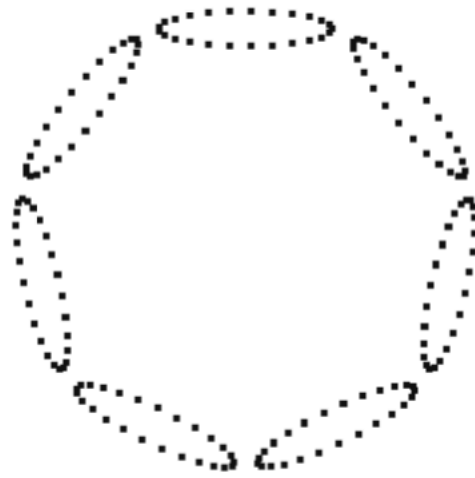
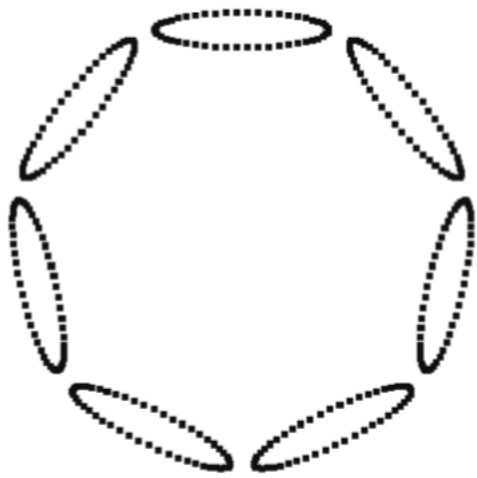


# Outline-Parameter-free denoising algorithm

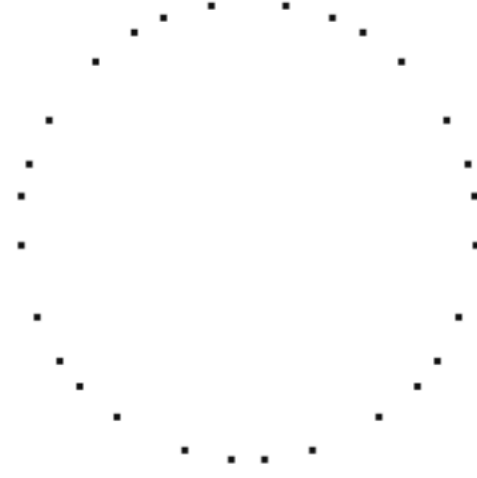
- Parameter-free denoising algorithm
  - Introduction
  - Sampling condition
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  - Discussions

# Declutter algorithm

- One parameter is needed



$k=2$



$k=10$

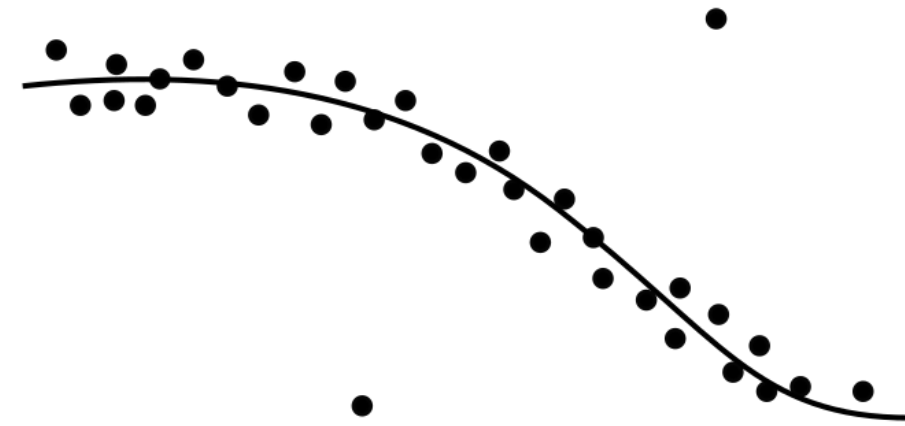
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**Data:** Point set  $P$ , parameter  $k$

**Result:** Denoised point set  $Q$

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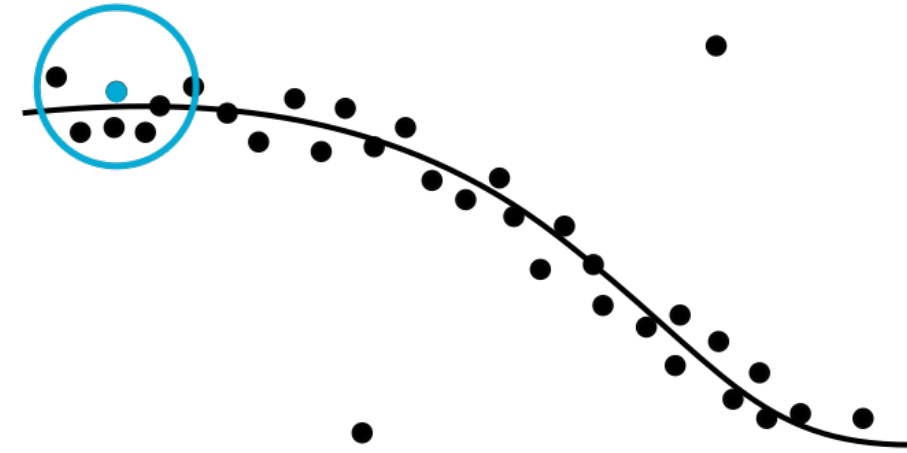
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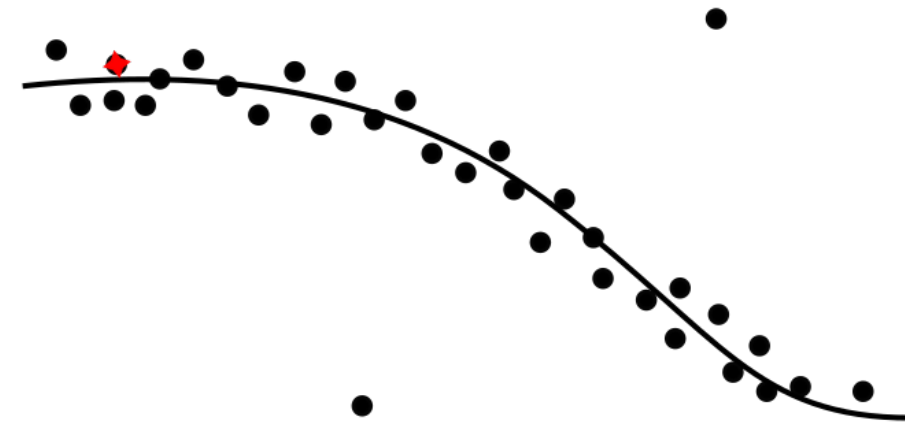
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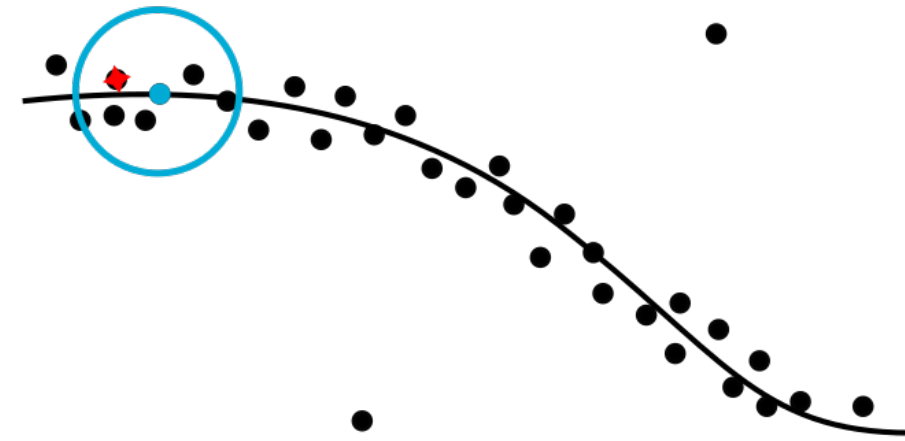
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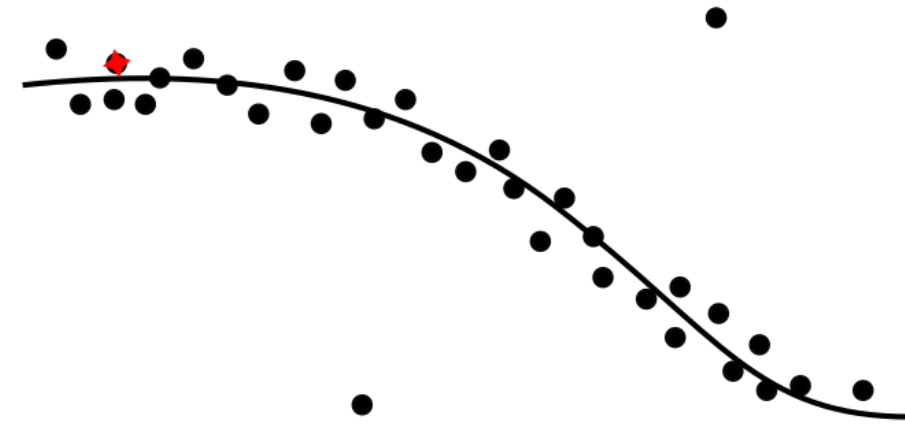
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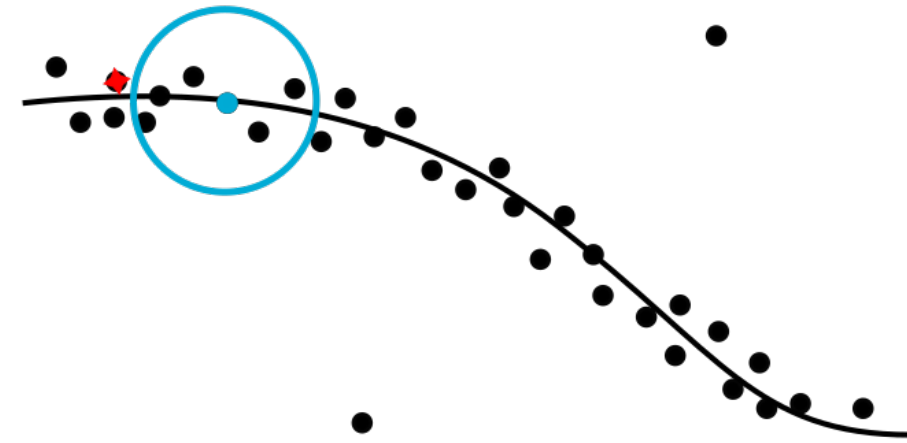
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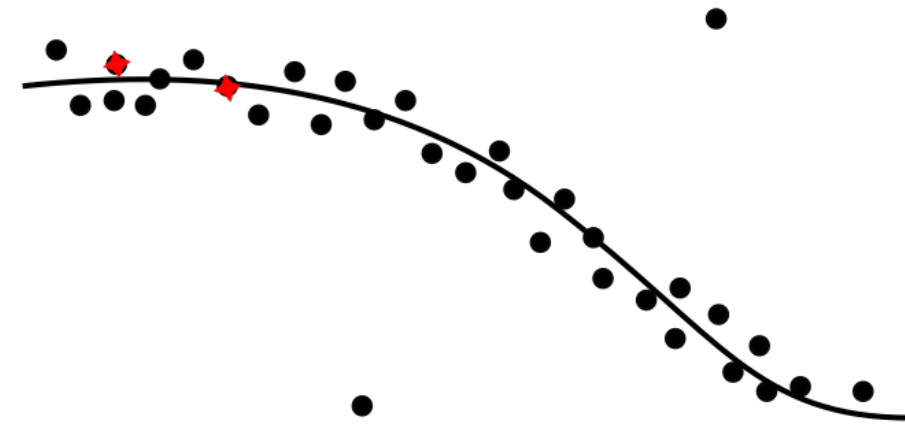
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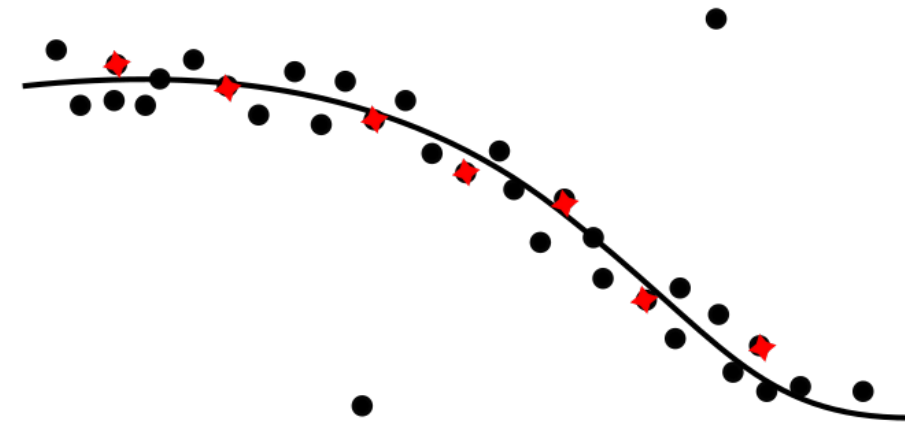
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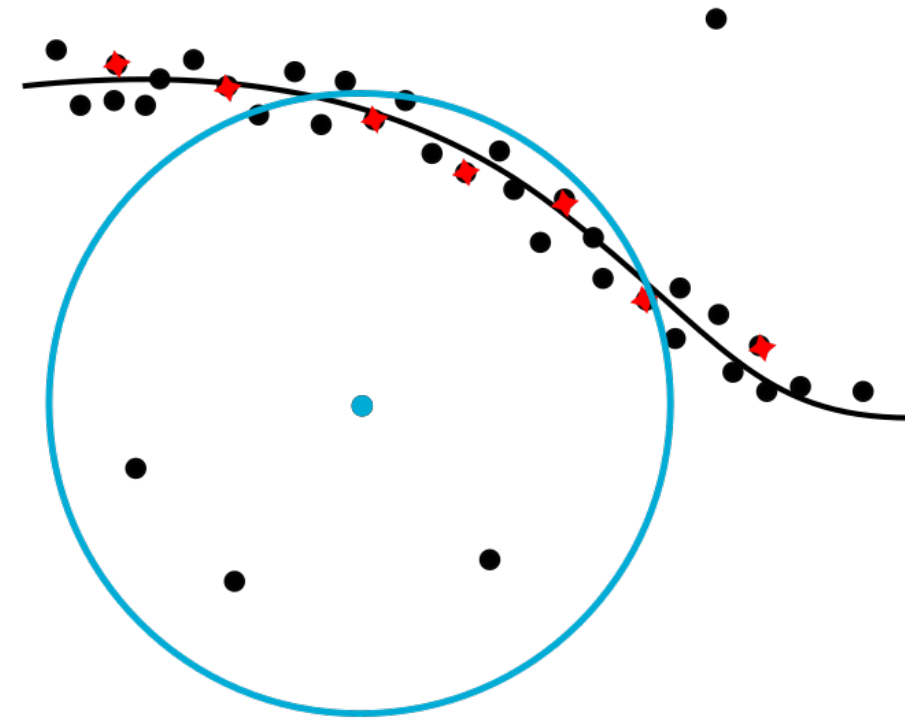
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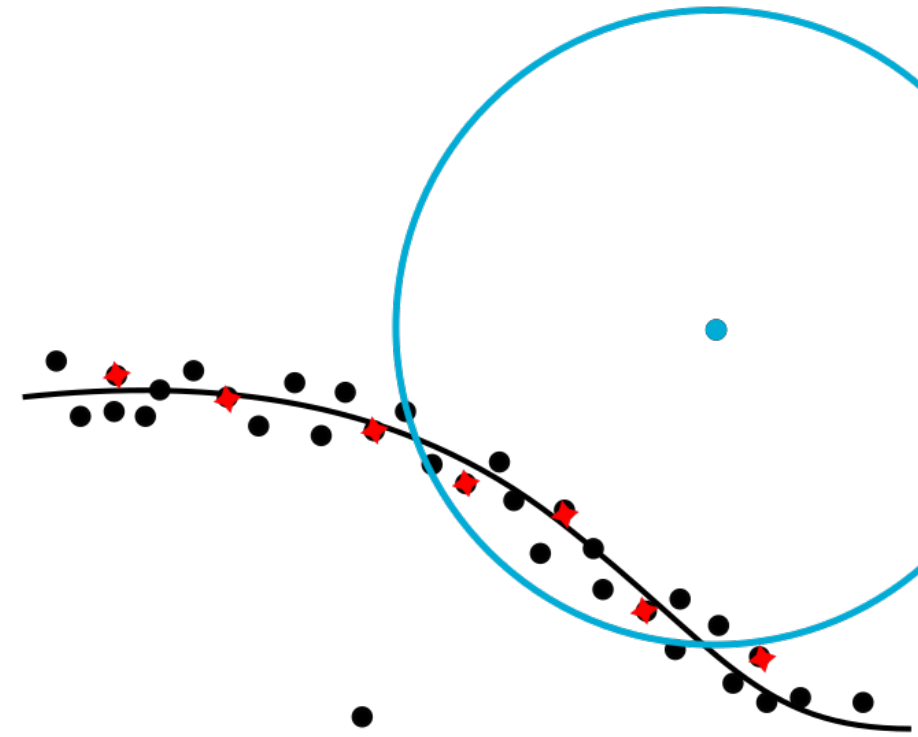
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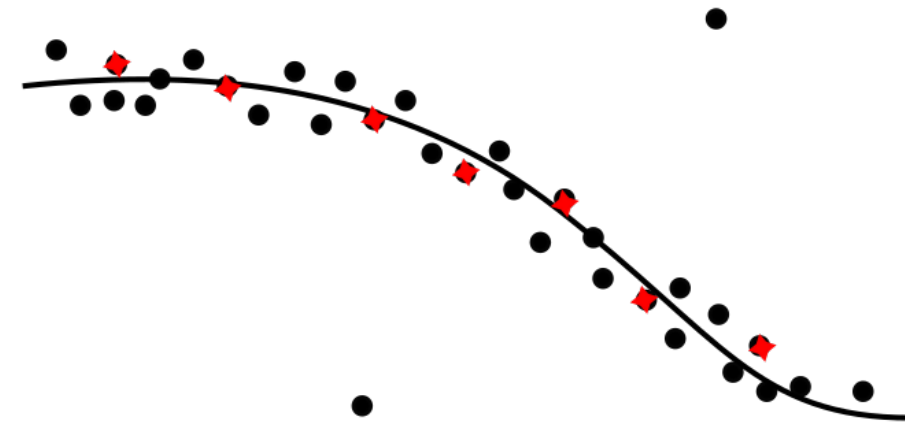
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# Declutter algorithm

**Theorem** *Given a point set  $P$  which is an  $\epsilon_K$ -noisy sample of a compact set  $K \subseteq \mathbb{X}$ , Algorithm Declutter returns a set  $Q \subseteq P$  such that  $\delta_H(K, Q) \leq 7\epsilon_K$ .*

*Hausdorff distance  $\delta(K, Q)$  between  $K$  and  $Q$*

*Infimum of  $\delta$  such that:*

$$\begin{aligned} \forall p \in Q, d_{\mathbb{X}}(p, K) &\leq \delta, \\ \forall x \in K, d_{\mathbb{X}}(x, Q) &\leq \delta \end{aligned}$$

# Declutter algorithm - adaptive version

**Theorem** *Given an  $\epsilon_k$ -adaptive noisy sample  $P$  of a compact set  $K \subseteq \mathbb{X}$  with feature size  $f$ , Algorithm Declutter returns a sample  $Q \subseteq P$  of  $K$  where  $\delta_H^f(Q, K) \leq 7\epsilon_k$ .*

Infimum of  $\delta$  such that:

$$\forall p \in Q, d_{\mathbb{X}}(p, K) \leq \delta f(\bar{p}),$$

$$\forall x \in K, d_{\mathbb{X}}(x, Q) \leq \delta f(x)$$

# Declutter algorithm – experiment

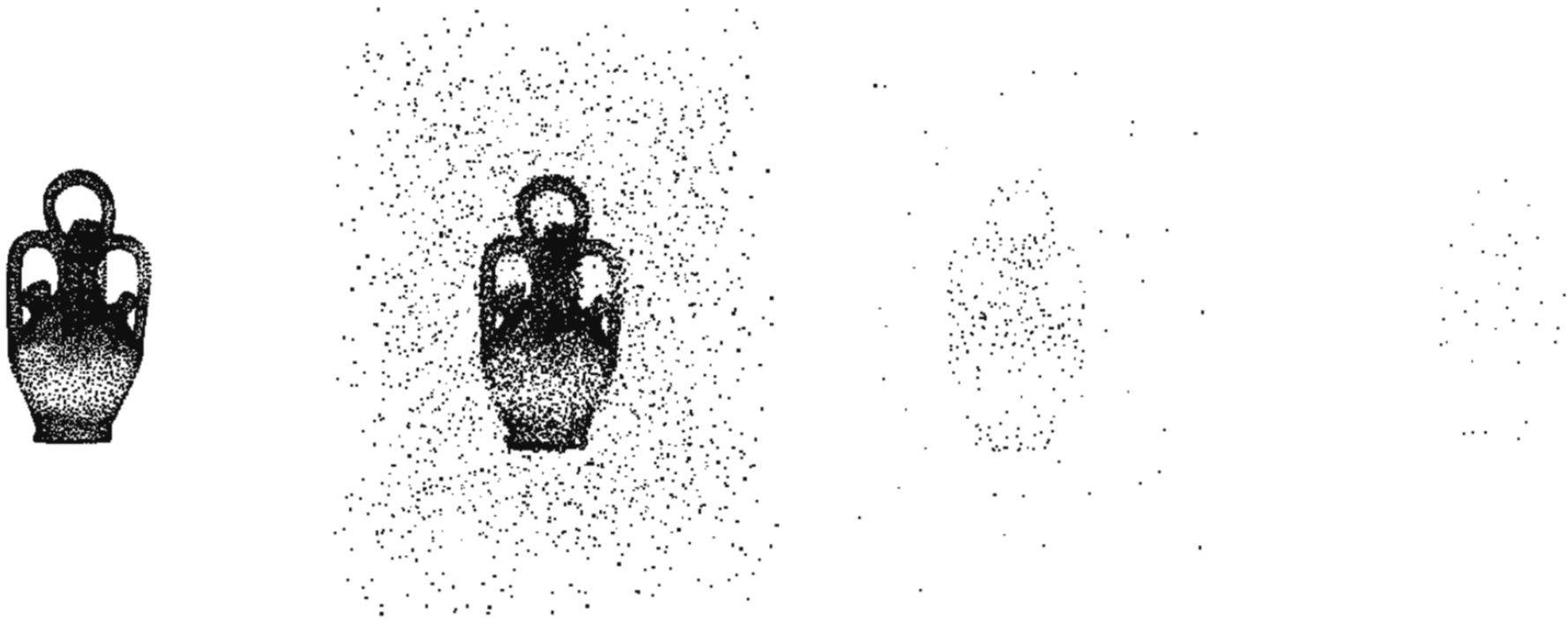


Figure 4: From left to right, the ground truth, the noisy input and the output of Algorithm Declutter for  $k = 81$  and  $k = 148$



# Outline-Parameter-free denoising algorithm

- Parameter-free denoising algorithm
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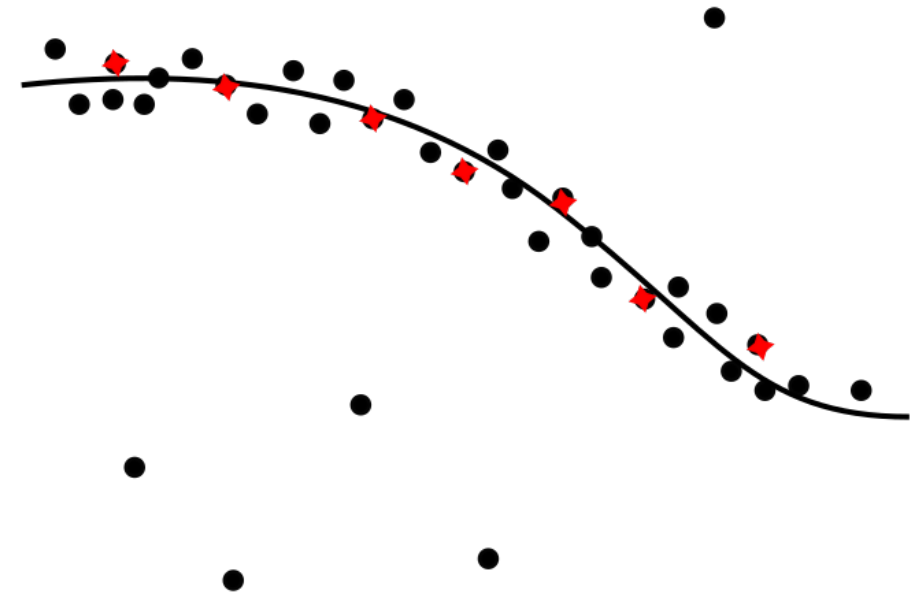
# Parameter-free algorithm

## Algorithm 2: ParfreeDeclutter( $P$ )

**Data:** Point set  $P$

**Result:** Denoised point set  $P_0$

```
1 begin
2   Set  $i_* = \lfloor \log_2(|P|) \rfloor$ , and  $P_{i_*} \leftarrow P$ 
3   for  $i \leftarrow i_*$  to 1 do
4      $Q \leftarrow \text{Declutter}(P_i, 2^i)$ 
5      $P_{i-1} \leftarrow \cup_{q \in Q} B(q, (10 + 2\sqrt{2})d_{P_i, 2^i}(q)) \cap P_i$ 
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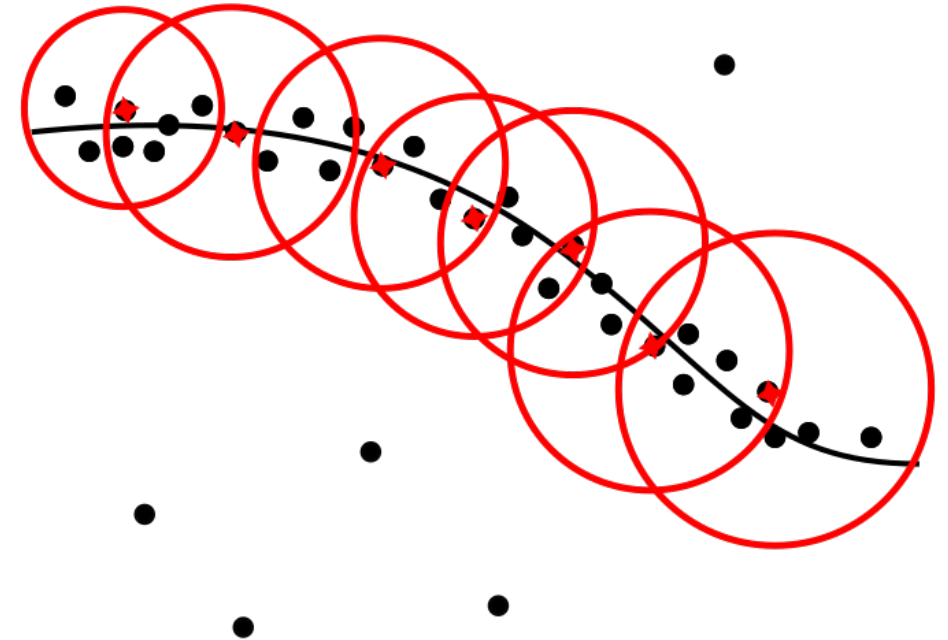
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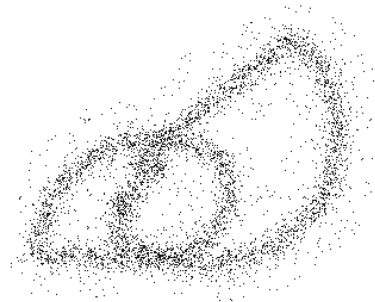
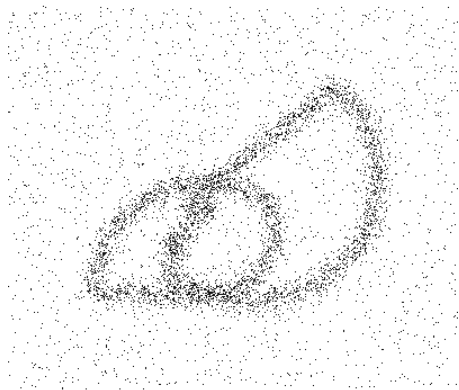
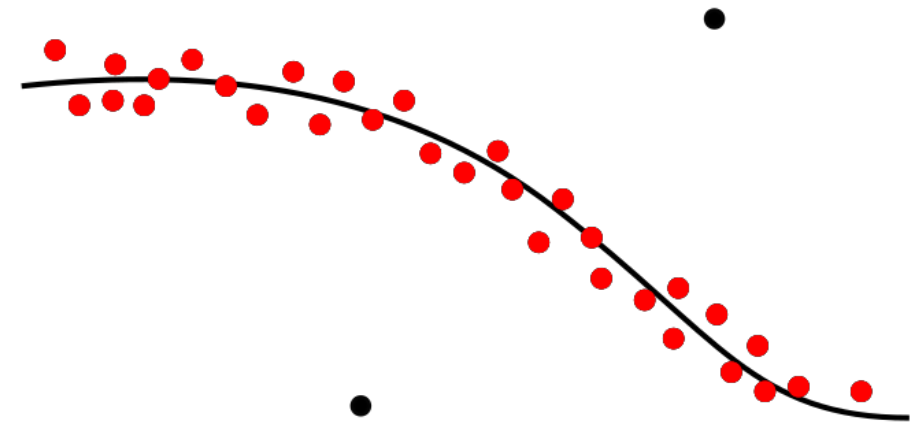
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```



k=128



k=64



k=2

# Parameter-free algorithm

**Theorem** *Given a point set  $P$  and  $i_0$  such that for all  $i > i_0$ ,  $P$  is a weak uniform  $(\epsilon_{2^i}, 2)$ -noisy sample of  $K$  and is also a uniform  $(\epsilon_{2^{i_0}}, 2)$ -noisy sample of  $K$ , algorithm ParfreeDeclutter returns a point set  $P_0 \subseteq P$  such that  $\delta_H(P_0, K) \leq (87 + 16\sqrt{2})\epsilon_{2^{i_0}}$ .*

- $P$  is a uniform  $(\epsilon_k, c)$ -noisy sample of  $K$  if:
  - $\forall x \in K, d_{P,k}(x) \leq \epsilon_k$
  - $\forall x \in \mathbb{X}, d_{\mathbb{X}}(x, K) \leq d_{P,k}(x) + \epsilon_k$
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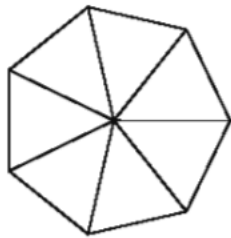
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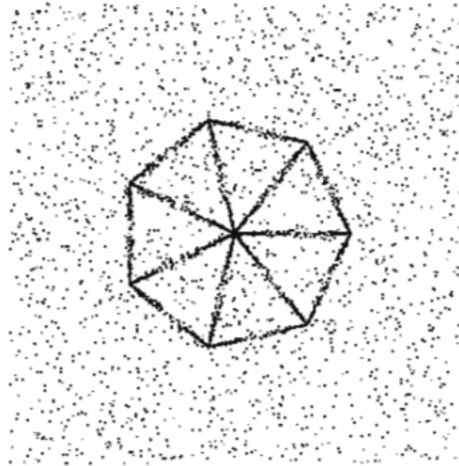
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  - $\forall p \in P, d_{P,k}(p) \geq \frac{\epsilon_k}{c}$

# Parameter-free algorithm

- Example where the parameter-free algorithm doesn't work.
- non-uniform



Ground truth



Noisy input



Intermediate step



Final output

# Parameter-free algorithm-experiment

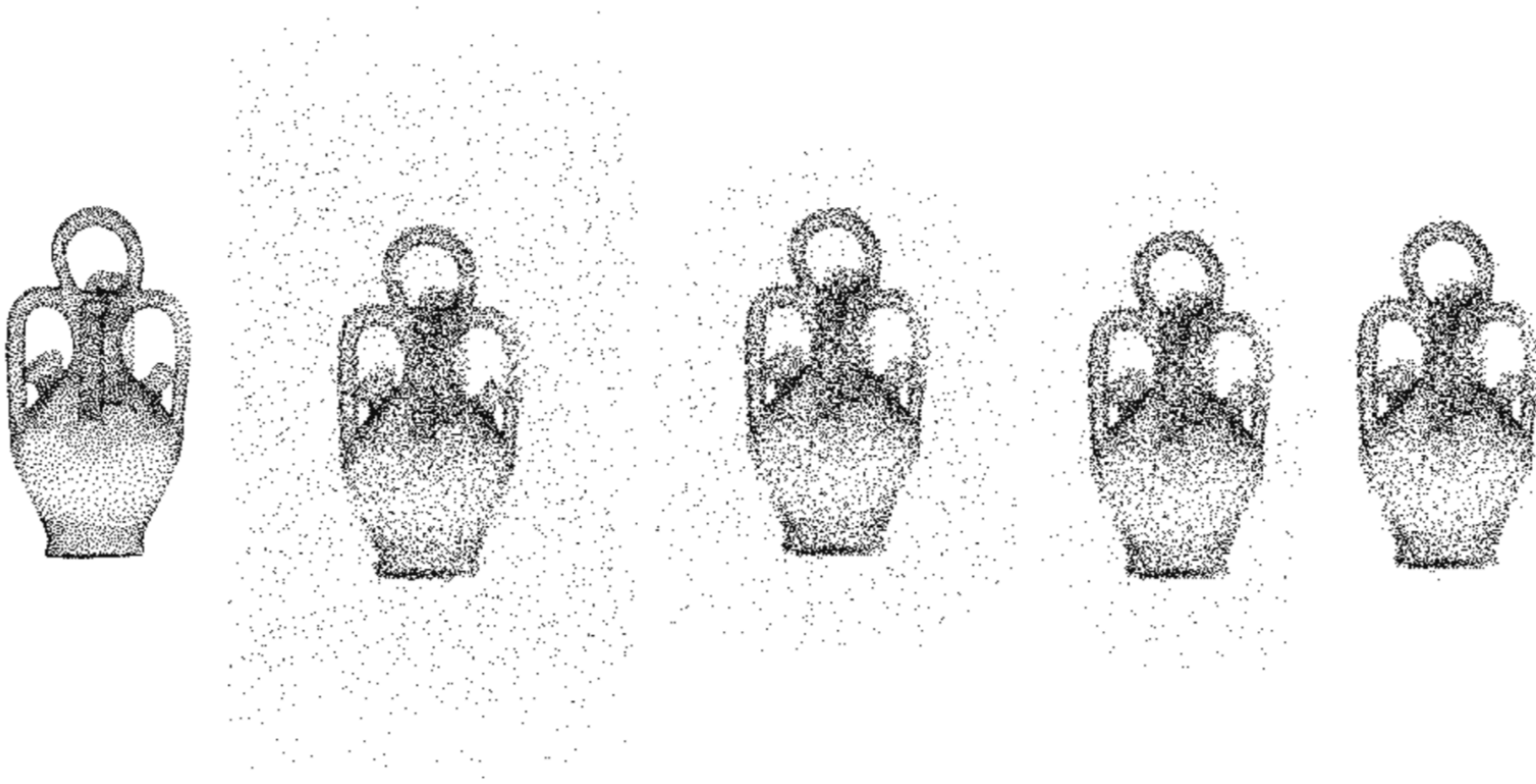
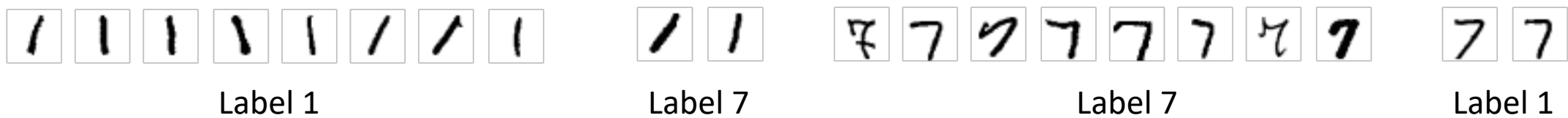


Figure 9: Experiment on a two dimensional manifold. From left to right, the ground truth, the noisy input, two intermediate steps of Algorithm ParfreeDeclutter and the final result.



# Parameter-free algorithm-experiment

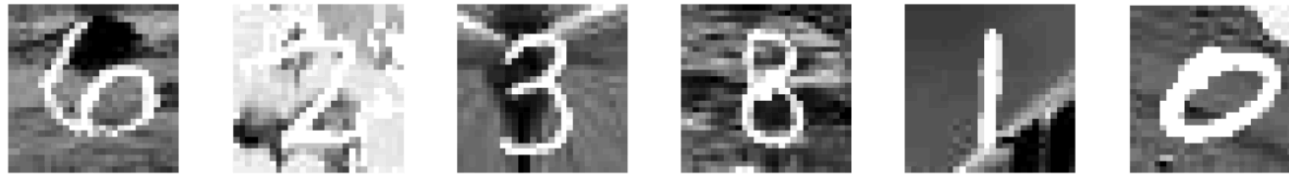
Swap noise:



1						Error(%)
2	Original	# Digit 1    1352		# Digit 7    1279		0.6564
3	Swap. Noise	# Mislabelled 1    270		# Mislabelled 7    266		4.0957
4		Digit 1		Digit 7		
5		# Removed	# True Noise	# Removed	# True Noise	
6	L1 Denoising	314	264	17	1	2.4500

# Parameter-free algorithm-experiment

background noise:



1					Error(%)	
2	Original	# Digit 1	1352	# Digit 7	1279	0.6564

7	Back. Noise	# Noisy 1 250		# Noisy 7 250		1.1464
8		Digit 1		Digit 7		
9		# Removed	# True Noise	# Removed	# True Noise	
10	L1 Denoising	294	250	277	250	0.7488

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# Discussions

- Relax the sampling conditions
- Estimate Hausdorff distance in a parameter free manner
- Adaptive theoretical guarantees for algorithm ParfreeDeclutter