Informed Novelty Detection in Sequential Data by Per-Cluster Modeling

Linara Adilova [1] linara.adilova@rub.de

Siming Chen [2] simingchen@fudan.edu.cn

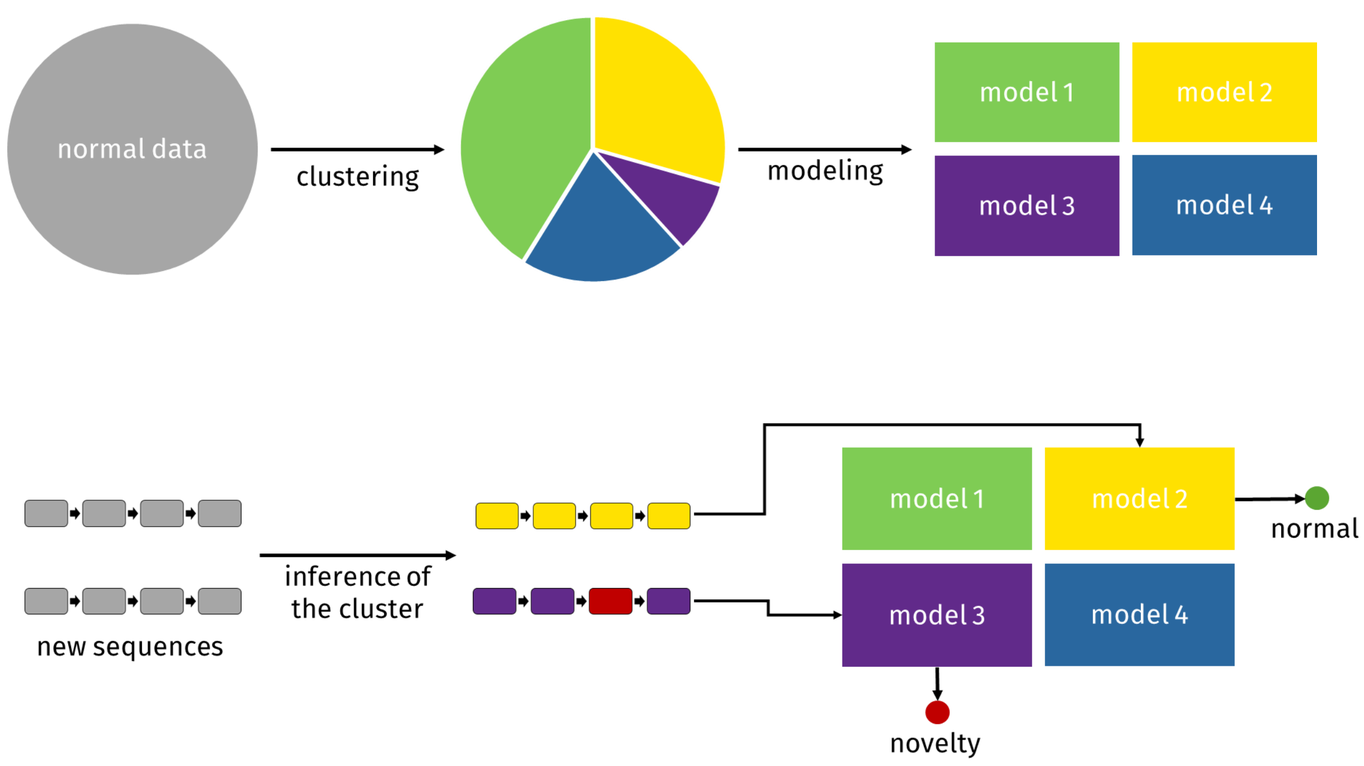
Michael Kamp [1,3,4] michael.kamp@uk-essen.de

[1] Ruhr University Bochum, Germany [2] Fudan University, China [3] Monash University, Australia [4] Institute for AI in Medicine (IKIM), Germany

Cluster first, then model data!

Human interaction outperforms automatic clustering: informed clustering by experts via visual analytics.

Novelty detection in discrete sequences



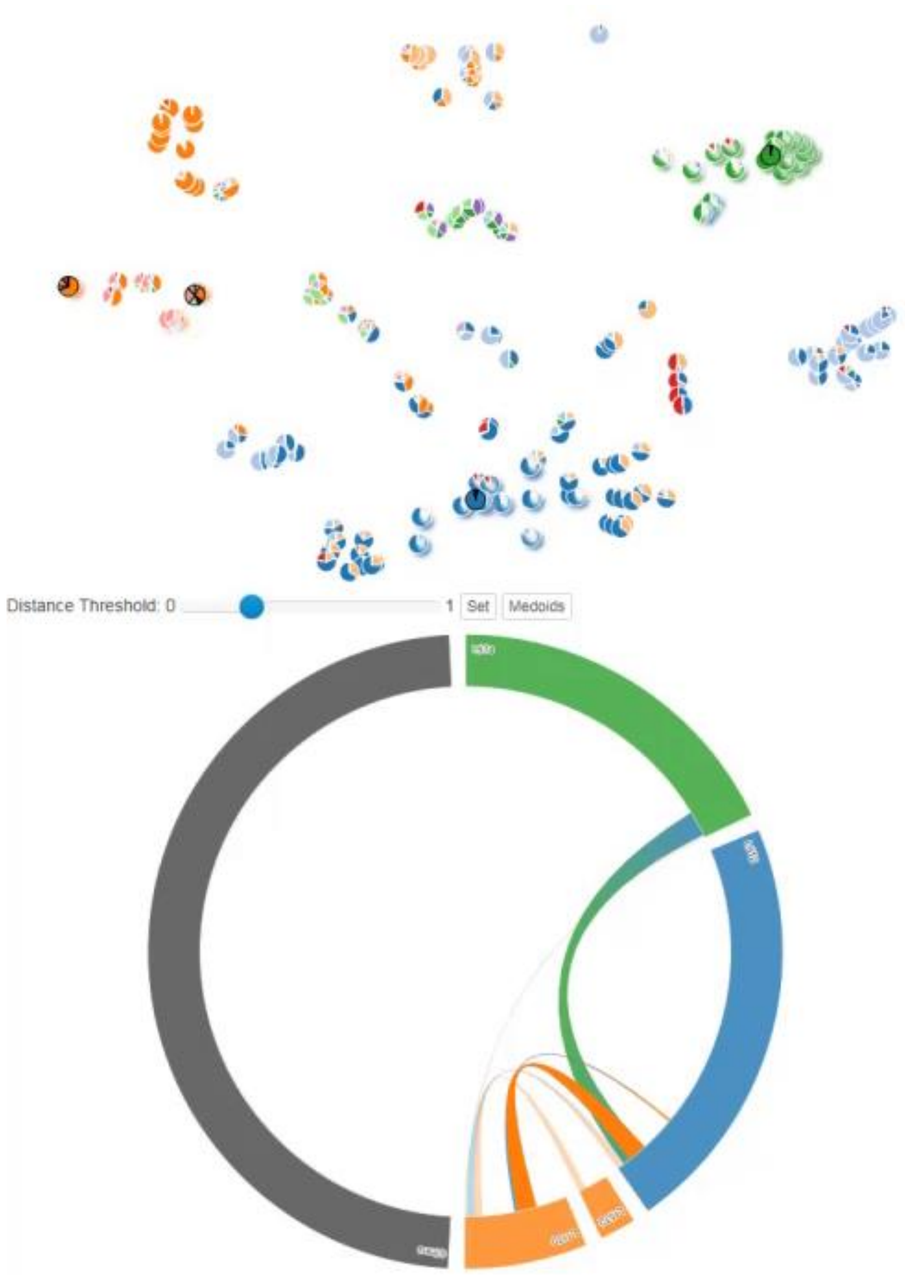
Algorithm 1 Novelty Detection via Per-Cluster Modelling

Input dataset $X \subset \mathcal{V}^*$, threshold $\theta \in \mathbb{R}$, sequence $s' \in \mathcal{V}^*$

Output $\{0, 1\}$ (0 for a normal sequence, 1 for a novelty)

Training:
obtain clustering C with k clusters of X
for $\mathcal{G}_i = \{s \in X \mid C(s) = i\}$ with $i = 1, \ldots, k$ **do**
  train process model h_i on $\mathcal{G}_i$
end for
Inference:
compute cluster $C(s')$ of s'
if $PP(h_{C(s')}, s') > \theta$ **then**
  return 1
else
  return 0
end if

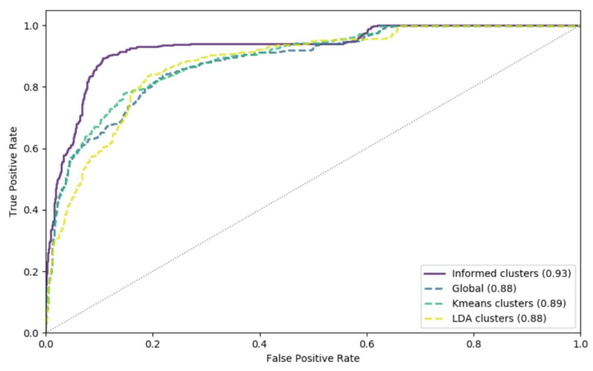
Expert-informed clustering

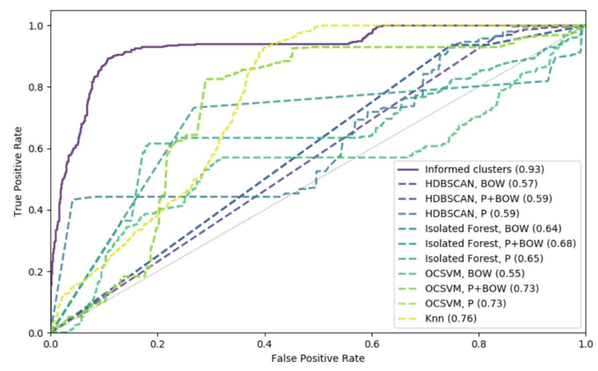


For a detailed demonstration see

Evaluation

Cybersecurity dataset





Method	AUC	$\frac{\text{Sens.}+\text{Spec.}}{2}$	Sens.	Spec.
IC-LSTMs	**0.93**	**0.89**	0.89	0.89
Global LSTM	0.88	0.81	0.80	0.81
k-means Cluster LSTMs	0.89	0.81	0.83	0.78
LDA Cluster LSTMs	0.88	0.82	0.81	0.82

Fake reviews dataset

Method	AUC	$\frac{\text{Sens.}+\text{Spec.}}{2}$	Sens.	Spec.
IC-LSTMs	**0.58**	**0.58**	0.58	0.57
Global LSTM	0.55	0.54	0.53	0.55
k-means Cluster LSTMs	0.52	0.51	0.50	0.52
LDA Cluster LSTMs	**0.58**	0.57	0.57	0.57

CPU utilization timeseries

Method	AUC	$\frac{\text{Sens.}+\text{Spec.}}{2}$	Sens.	Spec.
IC-LSTMs	**0.99**	**0.87**	0.77	0.97
Global LSTM	0.96	0.86	0.77	0.94
k-means Cluster LSTMs	0.97	0.84	0.85	0.82
LDA Cluster LSTMs	0.98	**0.87**	0.77	0.97