

Count What You Want: Exemplar Identification and Few-shot Counting of Human Actions in the Wild

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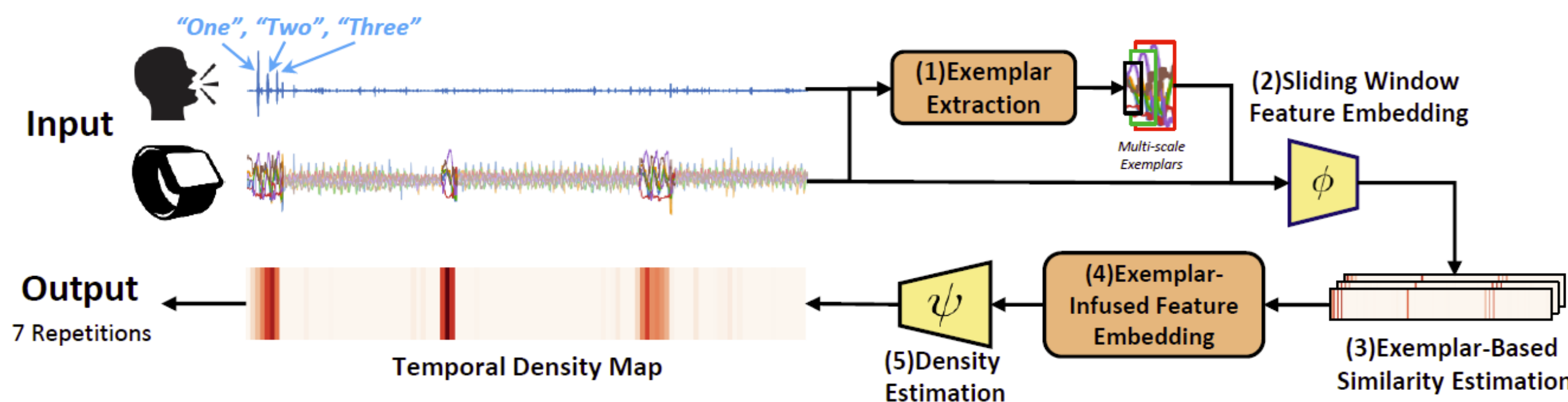


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Introduction

Our method focus on counting the number of repetitive actions with user-specified exemplars on wearable device. This exemplar is provided with audio cues (“one”, “two”, “three”).

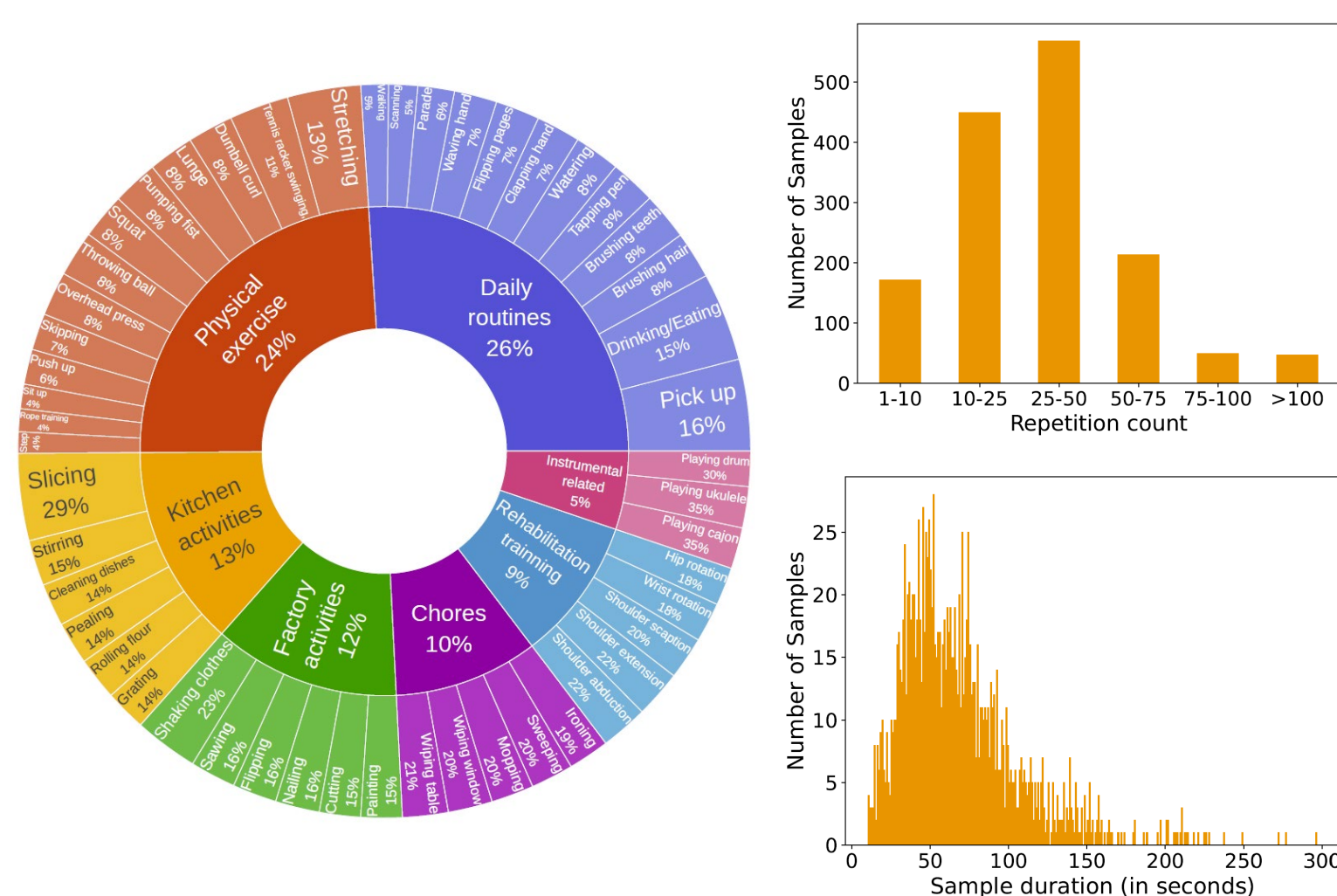


Motivation: Previous methodologies relying heavily on temporal self-similarity, encounter a significant limitation that they tend to focus solely on the most repetitive temporal patterns. This approach overlooks less frequent but potentially crucial actions.

Main Contributions:

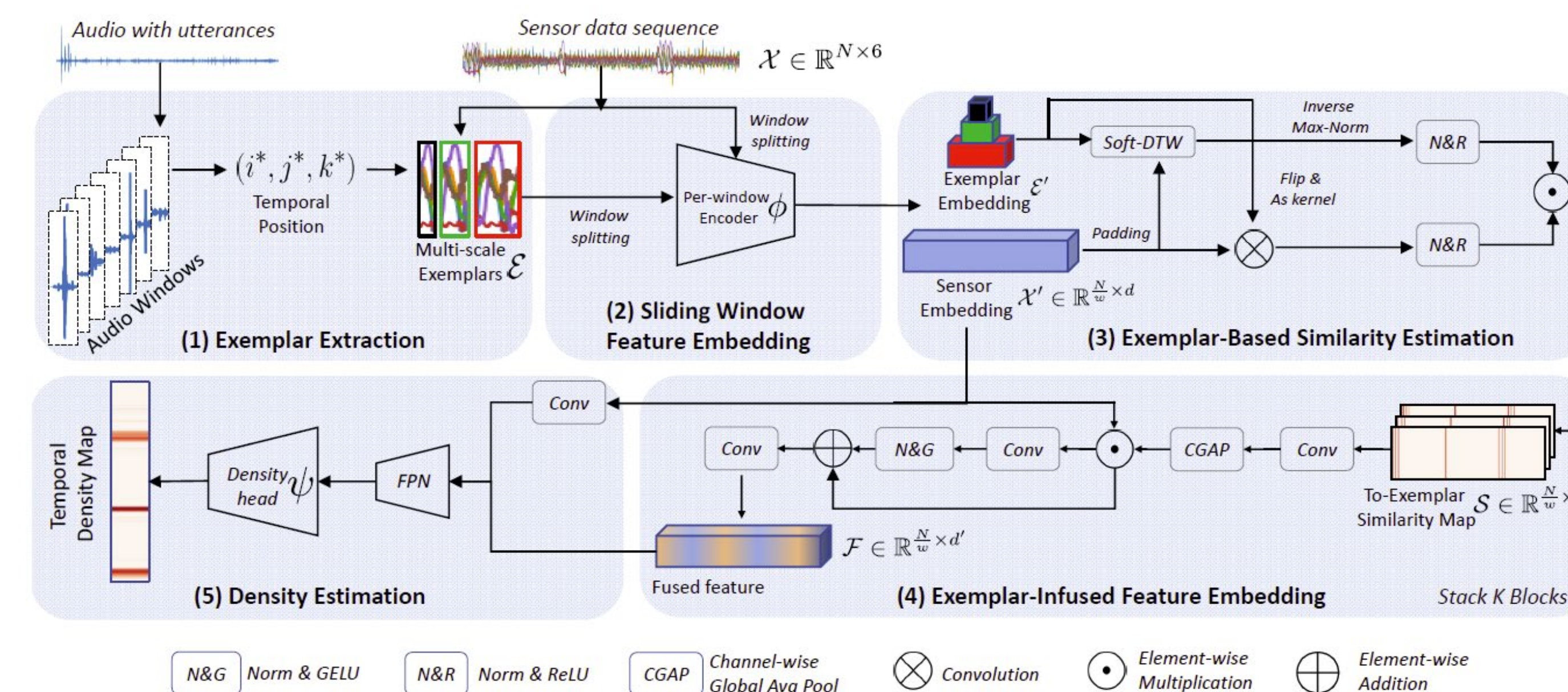
- ❑ A novel strategy for using audio prompts to specify exemplars of what needs to be counted.
- ❑ A novel counting method that utilizes exemplars, incorporating a distance-preserving loss and an exemplar-based data synthesis pipeline.
- ❑ An unique dataset with multiple data modalities to develop a practical counting method for real-world scenarios.

DWC Dataset



We introduce a new action counting dataset named DWC (Diverse Wearable Counting). This dataset, comprising 1502 entries from 37 subjects, spans seven categories — kitchen activities, household chores, physical exercises, factory tasks, daily routines, instrument-involved activities, and rehabilitation training. Encompassing 50 distinct action classes, DWC offers significantly greater diversity than existing datasets in this area.

Method



Pretraining with Synthesized Data :

- ❑ Fragment extraction with audio cues
- ❑ Augmentation with duration scaling, time shifting, amplitude scaling and random noise addition.

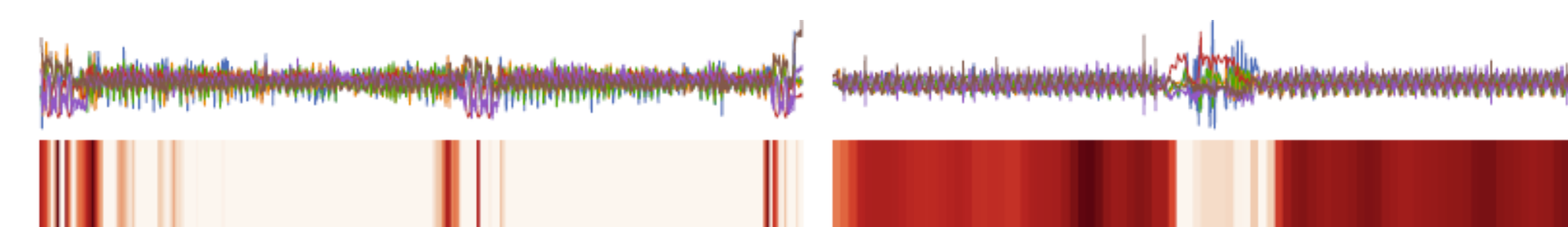
Training Objective :

- ❑ Counting Loss
 - ❑ Embedding Distance Preserving Loss
- $$(1) \mathcal{L}_c = (\text{sum}(\mathcal{T}) - \hat{c})^2$$
- $$(2) \mathcal{L}_{pl} = \mathcal{X}'^T \mathcal{L} \mathcal{X}' \mathcal{L} = \mathcal{D} - \mathcal{W}$$
- $$W_{ij} = \exp\left(-\frac{\|\mathcal{x}_i - \hat{\mathcal{x}}_j\|^2}{2\sigma^2}\right)$$

Result

Method	Val Set		Test Set	
	MAE	RMSE	MAE	RMSE
Mean	17.18	21.91	14.80	17.49
Frequency-based	28.10	45.31	28.65	45.39
RepNet	11.95	17.33	10.82	14.75
TransRAC	14.51	20.40	12.97	16.82
Proposed	7.66	12.25	7.47	13.09

Components	Combinations				
	✓	✓	✓	✓	✓
Pretrain	✗	✗	✗	✗	✓
Dist. Preserving Loss	✗	✗	✗	✓	✓
Constrained Detection	✗	✗	✓	✓	✓
Similarity Estimation	✗	✓	✓	✓	✓
MAE	11.30	10.87	10.32	10.05	7.66
RMSE	16.15	15.23	14.96	14.72	12.25



Code & Data

