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Enabling Ultra-low Power Machine Learning at the Edge

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## A framework for dataset construction including fused data from Human and Remotely Operated Vehicles (ROVs)

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#### Introduction

- Remotely Operated Vehicles (ROVs) → widely used in many critical missions.
- Human operator must control the ROV under stressful conditions and harsh environments → risk of unintentional movements → unwanted results.
- Need → control mechanisms embedded on the ROV to prevent accidents → easy implementation with the use of non-invasive wearable sensors.
- Monitoring the human through physiological signals along with an embedded mechanism controlling → helpful in search and rescue missions.
- Aim: a framework for dataset construction consisting of human and ROV data

## **Proposed Framework**

#### **Dataset Acquisition Setup**

- 6 subjects (5 males & 1 female), 23-29 years old, 3 experienced and 3 nonexperienced operators.
- 2 repetitions → stimulate the fatigue that occurs under stressful conditions.
- Pre-defined course → 4 basic simple movements (left, right, forward, backward)
  & pass the ROV around a sign.
- Data only from the right hand → focusing on on-air motion.
- 3-5 sec movement → 10 sec resting → flight duration: 10-12 min.
- sEMG sensor, HR Sensor & ROV Drone

#### **Stress Induction Technique**

- 2 repetitions: 1 stress-free & 1 stress-full
- Dataset of irritating sounds<sup>1</sup>.

#### State-Trait Anxiety Inventory (STAI) Evaluation

- Questionnaire with 40 questions:20 for the state & 20 for trait anxiety<sup>2</sup>.
- Filled in prior and at the end of the experiment.

#### **Data Fusion**

- Record of data = 8 sEMG values + 1 HR value + 11 values from the ROV = 20 signals.
- Annotated manually using video recordings from data collection process.



**Figure 4.** Setup for Data Collection

- 1. Wanlu Yang et al. "Affective auditory stimulus database: An expanded version of the International Affective Digitized Sounds (IADS-E)". In: Behavior Research Methods (2018), pp. 1415–1429.
- 2. Konstantinos N Fountoulakis et al. "Reliability and psychomet-ric properties of the Greek translation of the State-Trait Anxiety Inventory form Y: preliminary data". In: Annals of GeneralPsychiatry (2006), pp. 1–10.



## **Data Acquisition**

#### **sEMG Signal Acquisition**

- MYO Armband
- 8 active electrodes
- 200Hz sampling rate
- Bluetooth connection



Figure 1. MYO Armband

#### **HR Signal Acquisition**

- Polar OH1 Heart Sensor
- 1Hz sampling rate
- Bluetooth and ANT+ connectivity



Figure 2. Polar Sensor

#### **ROV's Signal Acquisition**

- DJI Spark Drone
- 200Hz sampling rate
- IMU's data



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Figure 3. DJI Spark Drone

## **Data Analysis**

#### **Feature Extraction**

- Window of 10 sequential time-frames
- 11 features for sEMG were evaluated → 7 were selected
- sEMG: VAR, WL, MAV, ZC, WAMP, RMS, SSC, MAX, MIN, SSI, IAV
- Raw HR
- ROV: Median, Mean, Min, Max

#### **Data Evaluation**

- Use of average values and box-plots
- Statistical t-test

Human Data								
	Raw	data	Features					
	Normal	Abnormal		Normal	Abnormal			
S1	-0.893	-0.887	VAR	82.079	29.944			
S2	-0.989	-0.975	WL	550.05	323.97			
S3	-0.974	-1.005	MAV	0.476	0.272			
<b>S4</b>	-0.9491	-0.951	ZC	9.375	4.237			
S5	-1.0347	-1.027	WAMP	9.909	3.742			
S6	-0.9276	-0.922	RMS	8.371	5.070			
<b>S7</b>	-0.8817	-0.887	SSC	8.835	3.680			
S8	-0.4045	-0.889	MAX	29.303	19.286			
			MIN	-29.695	-16.149			
HR	84.266	70.828	SSI	6676.2	2479.1			
			IAV	432.677	263.97			

Table 1. Average Values of human data

ROV Data								
	Raw dat	a	Features					
	Normal	Abnormal		Normal	Abnormal			
AccX	0. 003	0.038	Median	0.711	0.840			
AccY	-1.005	0.008	Mean	925.36	898.78			
AccZ	-1.03	-0.92	Max	-767.04	-725.42			
GyroX	0.382	0.361	Min	31.668	41.051			
GyroY	0.331	-0.492						
GyroZ	0.866	0.079						
MagX	85.430	190.434						
MagY	-741.67	-714.84						
MagZ	914.33	888.57						
Altitude	82.751	981.973						
R. Height	7.012	6.396						

Table 2. Average Values of ROV data



Human Data – Statistical T-test				ROV Data – Statistical T-test				
Raw		Features		Raw		Features		
	p < 0.05		p < 0.05		p < 0.05		p < 0.05	
S1	Х	VAR	٧	AccX	٧	Median	٧	
S2	х	WL	٧	AccY	√	Mean	√	
S3	х	MAV	٧	AccZ	√	Min	√	
S4	х	ZC	٧	GyroX	√	Max	√	
S5	х	WAMP	٧	GyroY	V			
S6	х	RMS	٧	GyroZ	x			
<b>S7</b>	х	SSC	٧	MagX	V			
S8	٧	MAX	٧	MagY	V			
		MIN	٧	MagZ	V			
		SSI	V	Altitude	V			
		IAV	V	Rel. Height	√			

Table 3. Statistical T-test values for Human and ROV

#### Results

- STAI did not show any significant difference prior and at the end of the experiment.
- Average values after extracting the features are different between normal and abnormal movements.
- Statistical analysis between normal and abnormal movements showed that the data are statistically different between the two movements.
- The selected features for sEMG and ROV are promising when used for classification.

#### **Conclusions**

- Propose a framework for the construction of a dataset including data from human and an ROV during a mission.
- Construct a dataset including stressfree and stressful data from human and an ROV, annotated manually.
- We perform data analysis which includes extracting the average values and performing a statistical t-test to conclude to the most suitable features.

#### **Future Work**

- Construct a larger dataset with operators with different degrees of experience.
- Include more physiological signals such as Electrodermal Activity (EDA), Accelerometer, Blood Volume Pulse (BVP) and Skin Temperature (ST).
- Develop classification algorithms used especially for anomaly classifications.
- Design of a shared-control mechanism for monitoring jointly the operator and the ROV and preventing the execution of commands given by involuntary movements.



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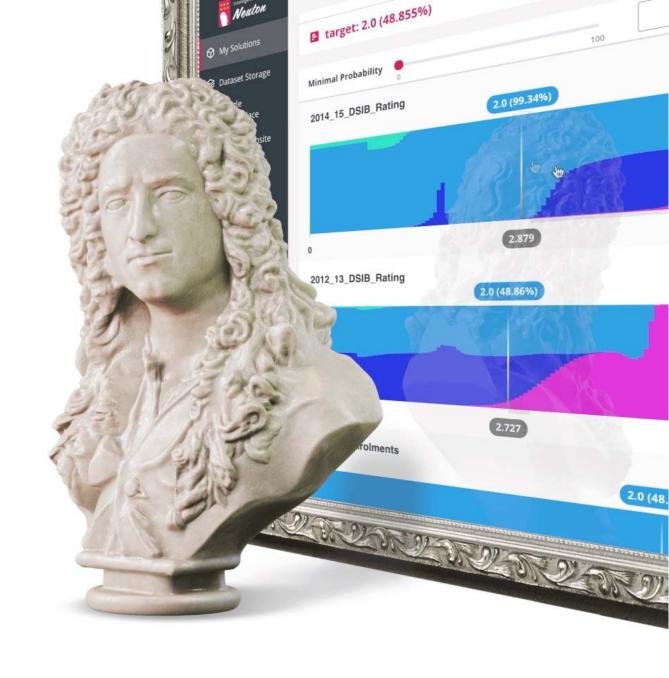
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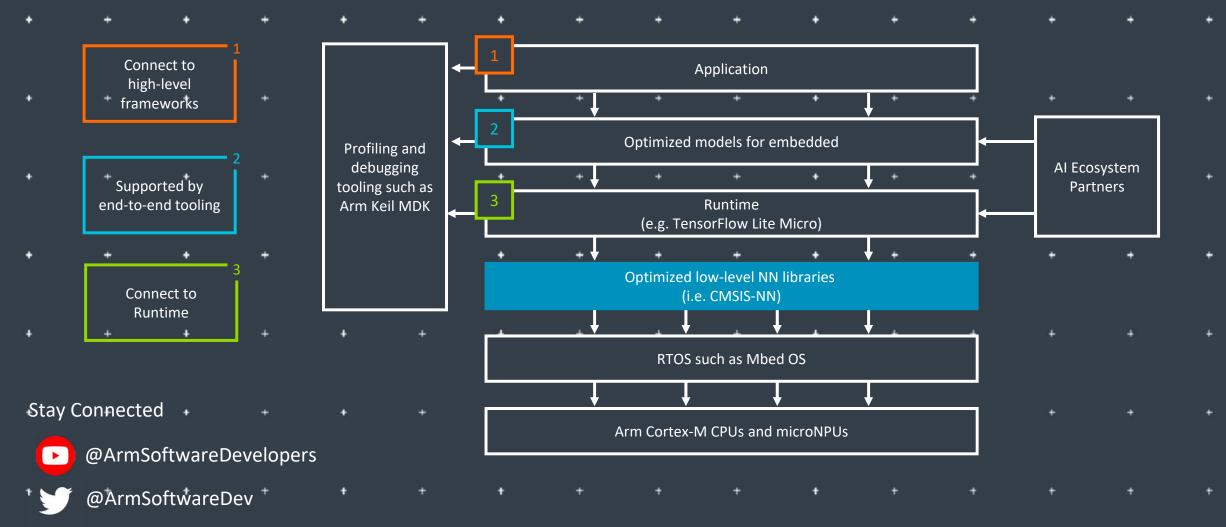
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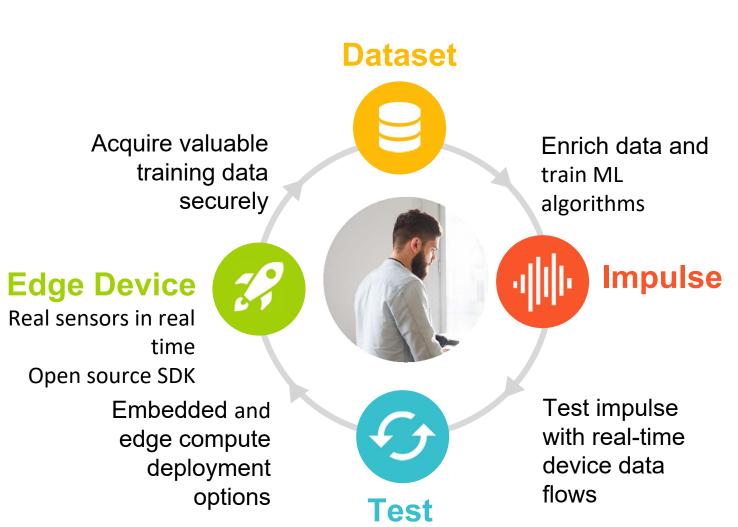


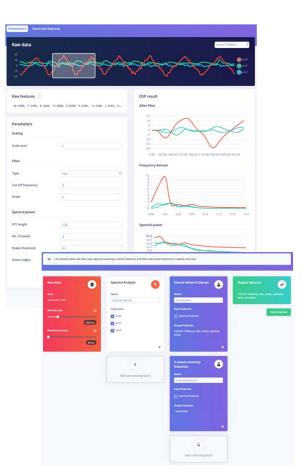
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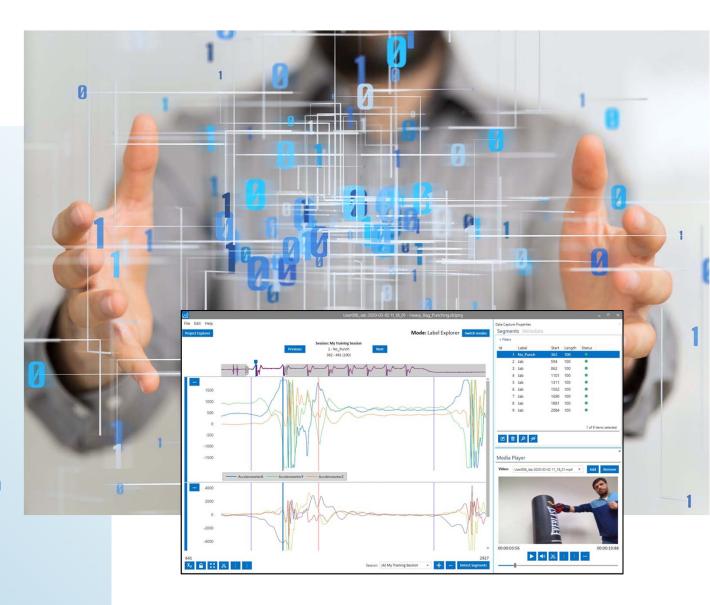


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