Developing Al Toolkits for the Remote Monitoring of Patients with PD, MS and Stroke - The ALAMEDA Project

Smart Diaspora 2023

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Project Motivation and Description





ALAMEDA - Bridging the Early Diagnosis and Treatment Gaps of Brain Diseases

- ▶ Digital transformation: a challenging necessity →rapid ageing of population; affects demands placed on the health care system
- ▶ Growing incidence of life-altering brain disorders →increased expenditure on health and long-term care
- ► ALAMEDA Vision: research and prototype the next generation of personalized Al healthcare support systems for people with brain diseases, specifically focusing on the needs of patients with Parkinson's, Multiple Sclerosis and Stroke (PMSS).



Why and what is ALAMEDA?

Research organisations























Healthcare organisations







HELLENIC REPUBLIC National and Kapodistrian University of Athens

SMEs and corporates















uni systems





Why and what is ALAMEDA?

Use cases

ALAMEDA

acknowledges that the care of patients with brain disorders is complex and manifestations of certain diseases could worsen over time and seriously impair the quality of life of patients and their caregivers: regular rehabilitation treatment assessments are essential to ensure that medical interventions are impactful and that the deterioration of state of health can be foreseen.

Digitally enhance rehabilitation treatment monitoring for stroke patients to provide continuous update on the patient recovery process.

Test a machine learning algorithm able to predict the risk of developing a relapse in Multiple Sclerosis.

Study the use of sensors in advanced
Parkinson's Disease to monitor motor and non-motor symptoms with the aim to predict the worsening of the manifestations of the desease.





Project Developments

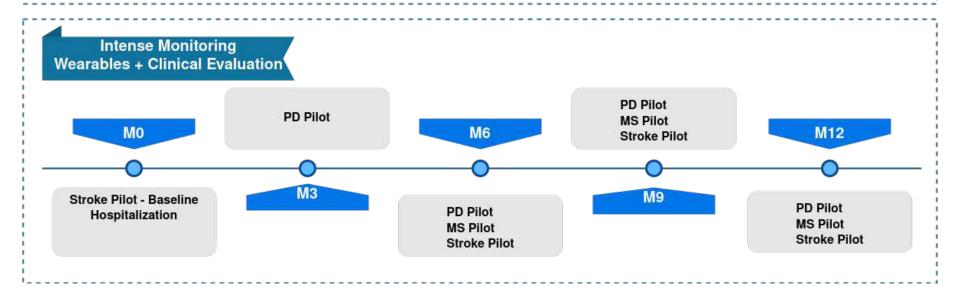




What we do in ALAMEDA - Patient Data Collection Journey

Continuous Monitoring

- Patient Reported Outcomes (PROs) Questionnaires and Relevant Questions from 5 health domains:
 - (i) Mobility, general motor or physical function, (ii) Sleep, (iii) Mental and cognitive ability, (iv) Emotional status, (v) Quality of Life and Daily Living
- Fitbit Smart Watch objectively monitored average activity and sleep metrics





What we do in ALAMEDA - Data Sources

The ALAMEDA technological ecosystem

Innovative data collection approach co-designed with patients, clinicians and caregivers



mHealth apps

The Digital Companion is ALAMEDA's application component that comprises of the **mobile applications** that the ALAMEDA patient



The WellMojo mobile application, developed by Wellics and customised for the ALAMEDA brain diseases' use cases, acts as a personal health coach, motivates, rewards and helps the users attain personal goals.

Chatbot

The chatbot android application developed by UNIC will provide a user-friendly graphical interface for patients and caregivers to interact with the ALAMEDA Conversational agent. Our aim is to offer a conversational interface mimicking human communication to personalize the patients' journey.

Mood Estimation Android App

Catalink is developing MEAA responsible for monitoring the user while s/he is interacting with other ALAMEDA core modules. MEAA's goal is to analyze the facial expressions of the users and assess their overall mood while they are conducting the related tasks.

Virtual Keyboard

The software, developed by CERTH, captures keystroke-related data as well as typing metadata, i.e., number of deletes, number of characters typed, typing session duration, deliberate long-press events, and the application where the user typed, while the content of the typed text is not recorded.

Virtual Supermarket Test Application

The VST is an app designed by CERTH to assess older adults' cognition through a simple task modeled on an everyday activity. The latest version of the VST includes advanced navigation metrics with the virtual space divided into three zones that represent different deviations.

Line tracking test application

The LTT is an app designed to assess older adults' hand dexterity. Developed within the NoTremor EU project, the Line Tracking Test measures the ability to follow a randomly moving target while ignoring the distracting target.

Smart devices

Multi-source data related to ALAMEDA use cases is collected from devices such as Fitbit, sleep mattress, smart insoles. smart bracelet and smart belt

Smart bracelet

This wearable allows to collect high-resolution raw acceleration data which are fused with raw data obtained from other specialized sensors to provide inputs for ALAMEDA algorithms. It also offers a variety of activity and sleep metrics using publicly available and validated algorithms

Sensorised insoles

Shoe pressure sensors are used to gather data that help assess the kinetics characteristics and fluctuations of ALAMEDA patients. They accurately measure the plantar force detected inside the shoe during all static and dynamic activities.

Smart watch/FitBit

smartwatch that monitors daily activities, heart rate data and possibly raw ration signals. The scope of this research is to enhance the provided sleep ssment by Fitbit. ICCS focuses on: sleep stages classification (Hypnograms estimation of sleep quality, monitor/assess the sleep properties along time.



Smart belt

he smart belt sensor, designed by NTNU, is a wearable device that is used to record patient motion activity during the day. One belt is composed of three motion tracking sensors mounted on a wearable belt for a patient's (user's) motion identification.



Smart mattress

A prototype mattress topper, that uses pressure sensors to model the patient's movement on the bed and produce heatmaps of the movement, has been developed by ENORA and distributed to the 3 pilots along with a commercial sleep analysis sensor.





his project has received funding from the European Union's Horizon 2020 research ovation programme under grant agreement No 101017558



What we do in ALAMEDA - Predictive Analysis Toolkits

- Semantic Knowledge Graph for Patient Status Modeling
- Movement and Mobility Analysis Toolkit
- Sentiment Analysis Toolkit
- Sleep Monitoring and Assessment Toolkit
- Predictor Variables Time Series Classification



What we do in ALAMEDA - Semantic Knowledge Graph

- Model collected data at a semantic level ALAMEDA Ontology
- ► Enables efficient querying and alert definitions

Variable	Description	Pilot Study	Alert Conditions
Domain I - Mobility, genera	al motor or physical function		
Step count, periods of relative immobility/slowness of movement	Continuously monitored step count and other features of general mobility in daily life	PD, MS, Stroke	Reduction of week average \geq 20%
Self-assessed questionnaire for muscle tone	Self-assessed questionnaire to quantify the muscle tone variable	Stroke	Increase of ≥ 1 point compared to previous month
ACTIVLIM questionaire	Self-assessed questionnaire to examine both upper and lower limb muscle strength using daily living activities	Stroke	When answer to a question changes from "easy" to "difficult" or "impossible", compared to previous month
Dizziness and Balance questionnaire	Self-assessed questionnaire for the balance variable	Stroke	When a patient checks a symptom that was not checked in the previous month
Domain II - Sleep disorders			
Pittsburgh Sleep Quality Index (PSQI)	Self-administered questionnaire to assess sleep patterns	PD, MS, Stroke	Change in answer to item 5, marking an increase in occurrence frequency compared to previous month
Domain III - Mental and co	ognitive ability		
Virtual Supermarket Test	Self administered test based on a 3D serious game to assess cognitive decline	PD	Deterioration of performance by \geq 20% compared to test 3 months ago
Domain IV - Emotional sta	tus		
PHQ-9	monitor the severity of depression and response to the treatment	MS, Stroke	Score increase of ≥ 1 point compared to previous month
Domain V - Quality of life	and daily living		
MFIS	Assessment of the effects of fatigue in terms of physical, cognitive and psychosocial functioning	PD, MS, Stroke	Score increase of \geq 16 points or \geq 19% compared to previous month



What we do in ALAMEDA - Semantic Knowledge Graph

CA output (JSON)

```
"user id": "abc123",
"source": "Conversational Agent",
"observations": [
    "sentiment scores": [
        "sentiment_class": "Positive",
        "sentiment score": 87.0
        "sentiment class": "Neutral",
        "sentiment score": 2.0
        "sentiment_class": "Negative",
        "sentiment score": 11.0
    "timestamp": "2022-05-17716:19:55Z",
    "explanation": "I feel great",
    "on dashboard": "false"
```

INSERT query (SPARQL)

```
INSERT DATA {
  :lef7f22d-2337-4e5d-9f53-359d6f9714f0 rdf:type :Patient ;
           :ID "abc123"^^xsd:string .
  :373c2ee0-6c27-4d0b-91e1-7ed34211fa4e rdf:type :Conversational Agent data ;
           :madeBySensor :ConversationalAgent ;
           :hasAgent :1ef7f22d-2337-4e5d-9f53-359d6f9714f0 ;
           :Message "I feel great" ;
           time:xsdDateTime "2022-85-17T16:19:55Z"^^xsd:dateTime ;
           on dashboard "false"^^xsd:boolean
  :373c2ee0-6c27-4d0b-91e1-7ed34211fa4e :hasResult :6836a00b-60bd-4501-b130-fca07bf76bac
            :6836a00b-60bd-4501-b130-fca07bf76bac rdf:type :Result ;
                       :hasScore 87.0 ;
                       :hasProperty :Positive .
  :373c2ee0-6c27-4d0b-91e1-7ed34211fa4e :hasResult :31762ea5-ada0-47e1-aa41-b034ede75b3b .
            :31762ea5-ada0-47e1-aa41-b034ede75b3b rdf:tvpe :Result :
                       :hasScore 2.0 ;
                       :hasProperty :Neutral .
  :373c2ee0-6c27-4d0b-91e1-7ed34211fa4e :hasResult :262dc52b-af86-4dc4-b7bc-2b3b62e4fc33 :
           :262dc52b-af86-4dc4-b7bc-2b3b62e4fc33 rdf:type :Result ;
    Semantic Representation (RDF triples)
                                                                                   -"abc123"
                                           "2022-05-17T16:20:43Z"^^xsd:dateTime
  Conversational Agent data
                         Conversational Agent O
                                                    time:xsdDateTime
                                                                       user_01
                                                                                     19.0"^^xsd:decima
                                                               hasAgent
                                             madeBySensor
   MoodMonitoringProperty
                                                                           hasScore
                                                 output 01
                                                              hasResult
  Patient
                                             Message
                                                                       has Property >
                                                  hasResult
                                                                    result 01
                              "I feel great!"
                                                         hasResult
  Result
                                                                     "22.0"^^xsd:decimal
                       "59.0"^^xsd:decimal <- hasScore —  result_03
                                                                hasScore
                                          hasProperty
  Sensor
                                                                    hasProperty-
                                                            result 02
                                                                              Negative
                                        Positive
```



What we do in ALAMEDA - Sentiment Analysis

- ► Multimodal, multi-tool approach
 - On-device app: Designed & implemented a simple mood estimation algorithm which runs only locally, on the smartphone
 →detect positive, negative and neutral mood
- **▶** Emotion Recognition Service:

 - - PD: Happy, sad, neutral, pain, surprise
 - MS: Happy, sad, fear, disgust
 - Stroke: happy, sad, neutral, pain
- ► Text-based sentiment analysis service:
 - Extract mood of patient based on free text interactions with the ALAMEDA Conversational Agent (classify utterances up to 150 tokens in length as positive, negative or neutral)



What we do in ALAMEDA - Movement and Mobility Analysis

"Umbrella" term for services that use raw data from wearables (bracelet, IMU belt, insoles) to detect

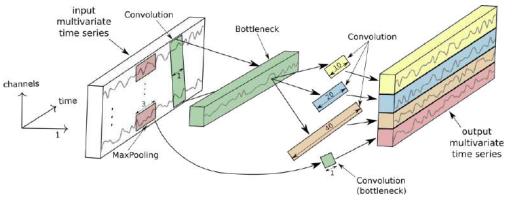
Current Work	Partner
Detect simple ADLs (walk, sit, stand, walk up/downstairs)	NTNU
 Detect tremor/dyskinesia/bradykinesia in PD Current experiments on Levodopa Response Study dataset, EMA Dataset, mPower Dataset 	ICCS, NTNU
Detect freezing of gait in PD - Current experiments on Gait in PD dataset	CERTH, NTNU
Detect execution of Stroke Rehabilitation Exercises based on accelerometer bracelet - Current experiments based on SPARS9X dataset for shoulder exercises	UPB



What we do in ALAMEDA - Movement and Mobility Analysis - Freezing of Gait in PD

► 1D-conv models based on custom *InceptionTime* [1] modules and Squeeze and Excitation Blocks

Daphnet Dataset	Sensitivit y	Specificity	F1 Score
InSEption	91.2%	94.1%	0.862
LN- Inception	85%	98.4%	0.856
CNN-LSTM	33.8%	97.1%	0.363
LSTM	71.3%	97.0%	0.727
Bi-RNN	83.6%	98.1%	0.842
CNN	25.6%	99.4%	0.285



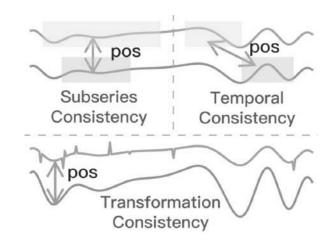
[1] Fawaz et al, 2019: InceptionTime: Finding AlexNet for Time Series Classification

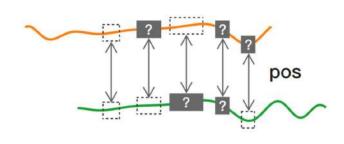
IMU Dataset	Sensitiv ity	Specific ity	Macro F1 Score
InSEpti on	98%	99%	0.972
LN- Inceptio n	97%	99%	0.962



What we do in ALAMEDA - Movement and Mobility Analysis - Timeseries representation Learning

- Time series representation learning using contrastive learning techniques → TS2Vec Model [1]
- Uses **contextual** consistency: representations at the same timestamp in two augmented contexts are positive pairs
- Context is generated by applying timestamp masking and random cropping on the input time series.
- Timestamp masking randomly masks the timestamps of an instance, and random cropping randomly samples overlapping time segments.





[1] Yue, Z. et al. Ts2vec: Towards universal representation of time series. AAAI 2022



What we do in ALAMEDA - Movement and Mobility Analysis - Timeseries representation Learning

- **Experimentation pipeline**

 - Use learned representations for "anomaly" detection:
 - Normal vs abnormal gait (insoles)
 - Daily Activities vs Rehabilitation Exercise sessions (accelerometer bracelets)
 - Study *generalization capability* of representations
 - How many samples from an activity or a specific exercise are required to discriminate between activity / exercise classes
 - What is the degree of inter-subject generalizability
 - Experiment with adversarial methods to discriminate the user who generated a time series



What we do in ALAMEDA - Sleep Monitoring and Assessment

Current Work	Partner
Detect Sleep Stages (light, deep, REM, awake) based on Wearable data from Smart Watch and Withings Sleep Mattress - Experiments performed on Apple Watch[1], MESA[2] and SHHS2[3] datasets	ICCS
 Sleep Quality Assessment Prototype 8x16 pressure sensor mattress topper to record body posture during sleep - correlate with heart rate, wake up events, duration of sleep stages 	ENORA

^[1] Olivia Walch, Yitong Huang, Daniel Forger, Cathy Goldstein, Sleep stage prediction with raw acceleration and photoplethysmography heart rate data derived from a consumer wearable device, Sleep, Volume 42, Issue 12, December 2019, zsz180.

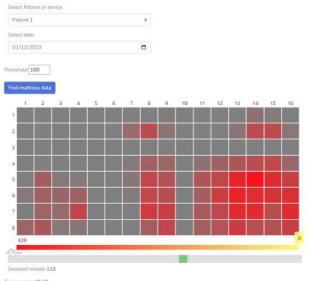
^[2] Chen X, Wang R, Zee P, Lutsey PL, Javaheri S, Alcántara C, Jackson CL, Williams MA, Redline S. Racial/Ethnic Differences in Sleep Disturbances: The Multi-Ethnic Study of Atherosclerosis (MESA). Sleep. 2015 Jun 1;38(6):877-88. doi: 10.5665/sleep.4732. PMID: 25409106; PMCID: PMC4434554.

^[3] Quan SF, Howard BV, Iber C, Kiley JP, Nieto FJ, O'Connor GT, Rapoport DM, Redline S, Robbins J, Samet JM, Wahl PW. The Sleep Heart Health Study: design, rationale, and methods. Sleep. 1997 Dec;20(12):1077-85. PMID: 9493915.



What we do in ALAMEDA - Sleep Monitoring and Assessment





Sleep Assessment using ENORA mattress topper

- Raw data acquired every 1 minute
- Posture results: Type, duration and rate of change of posture
- 13k readings so far
- End goal: define a sleep quality score and relate it to PD MDS-UPDRS scores



What we do in ALAMEDA - Variable Time Series Classification Service

"Umbrella" term for models that connect **all sources** to the **current** or **future patient health status**, expressed in terms of:

► MDS-UPDRS (for PD), EDSS (for MS), mRC (for Stroke)

Current Work	Partner
Wrapper service for model invocation	UPB
- Current experiments on PPMI, LRRK2 and Fox Insight Datasets	CERTH
 Estimate MS EDSS scores from PRO data Current experiments on MSOAC and Performance of ADL in PwMS datasets 	CERTH
- Starting experiments with PROMOPRO-MS dataset	UPB, ICCS

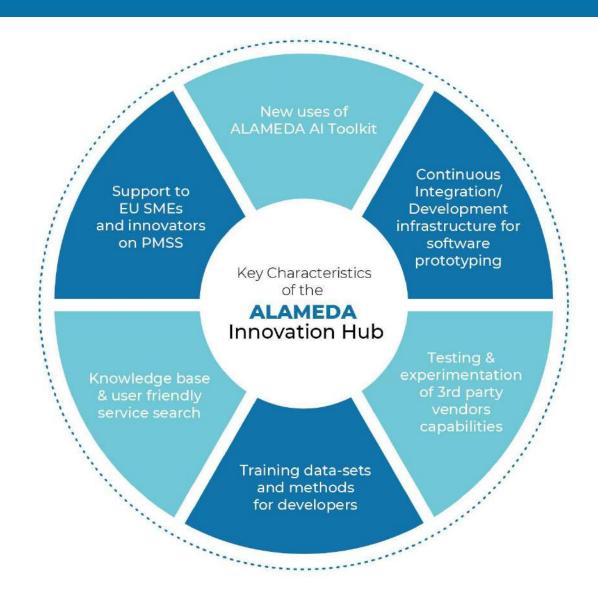


ALAMEDA - Project Sustainability





ALAMEDA - What Happens After





THANK YOU!