

# Deep Learning for Predicting Disease Progression of Clinical Endpoints in ALS

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### Motivation

- Amyotrophic Lateral Sclerosis (ALS) is a neurodegenerative disease.
- No cure and a short life expectancy following the start of symptoms.
- Most patients die of respiratory failure within 3-5 years.

### Problem

- ALS affects most muscles of the body.
- Several clinical conditions derive from it.
- Non-Invasive Ventilation showed the most promise in terms of extending life expectancy and improving quality of life.

### Objective

Our **main objective** is to increase life expectancy and improve quality of life. For that we propose the use of Deep Learning to:

- Predict ALS disease progression within a predefined time window (90, 180 and 365 Days).

### Background

- Prognostic models based on predefined time windows to predict the disease progression and the need of NIV treatment in ALS.
- Categorization of patients depending on their stage of disease progression, stratifying them in 3 groups, slow, neutral and fast.
- Use of neural networks (mainly LSTM) to predict the disease progression of the patients in the various time windows.
- Shapley Additive Explanations (SHAP) explaining the deep model was helpful to determine the influence of certain features for the overall prediction.

### Data

- Tabular Data from the ALS clinic of Centro Hospitalar Universitário de Lisboa Norte, Hospital Santa Maria [1].
- Static information
  - Demographic information about the patient.
- Temporal information – Multivariate Time Series
  - Results of a series of clinical assessments of specific tests:
    - Respiratory tests;
    - Neurophysiological data;
    - Questionnaire answers to the ALS Functional Rating Scale.
- For each of the following clinical endpoints, there is predefined time windows (90, 180 and 365 days).

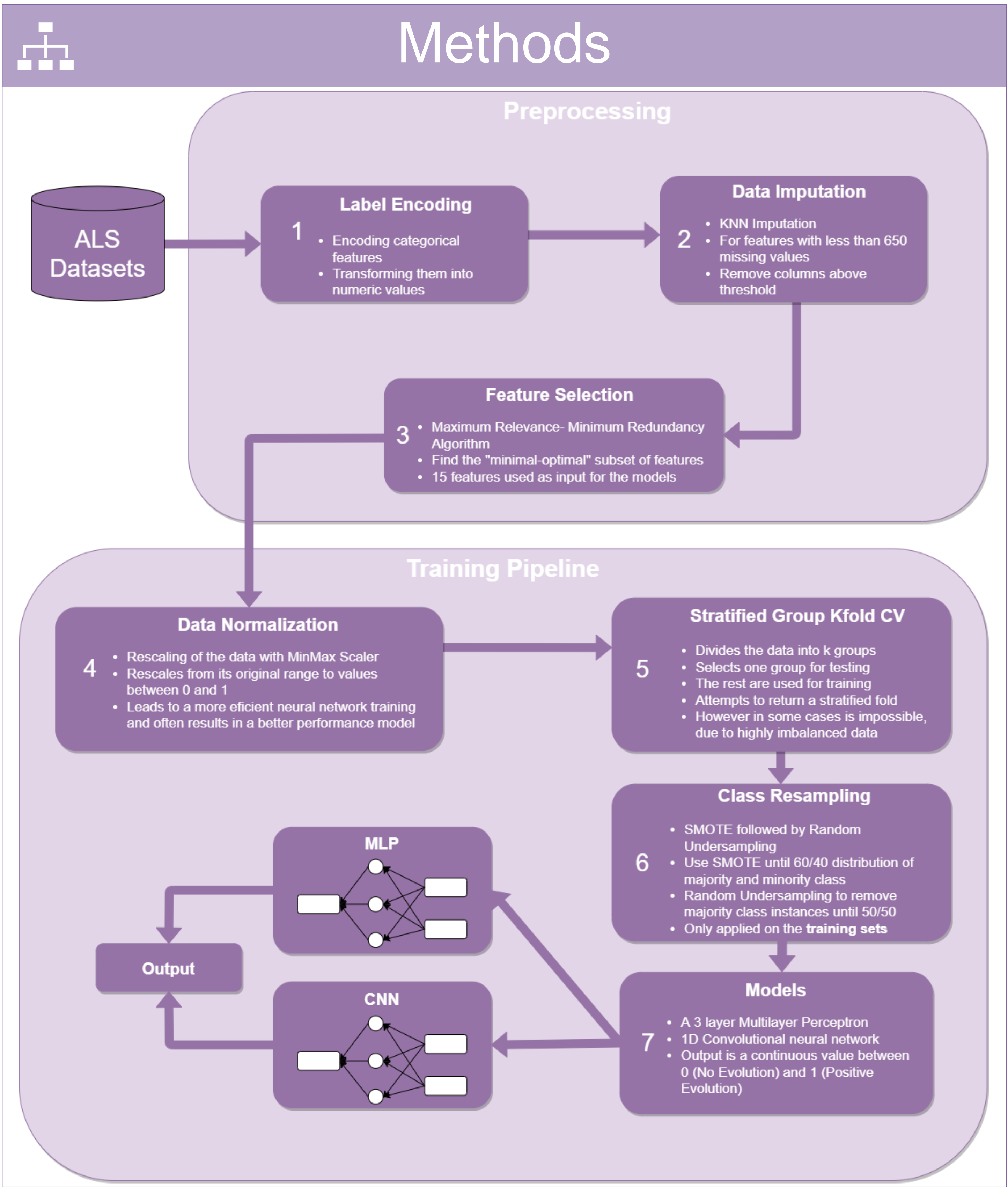
NIV

Aux Comm Device

PEG

Caregiver

Wheelchair



### Class Distribution

Dataset	Evolution	# Patients for the Time Windows in k Days		
		k=90	k=180	k=365
C1 – Need for NIV	Yes	217	740	1394
	No	4178	3655	3001
C2 – Need for na Auxiliary Communication Device	Yes	159	454	859
	No	5237	4942	4537
C3 – Need for PEG	Yes	68	267	605
	No	5983	5784	5446
C4 – Need for a Caregiver	Yes	341	837	1437
	No	2750	2254	1654
C5 – Need for a Wheelchair	Yes	202	589	1242
	No	4928	4541	3888

### Results

Dataset	Model	AUC ROC Mean		
		k=90	k=180	k=365
C1 – Need for NIV	MLP	0.761	0.746	<b>0.792</b>
	CNN	0.768	0.766	0.762
C2 – Need for na Auxiliary Communication Device	MLP	X	<b>0.924</b>	<b>0.924</b>
	CNN	X	0.921	0.906
C3 – Need for PEG	MLP	X	X	X
	CNN	X	X	X
C4 – Need for a Caregiver	MLP	0.834	0.797	0.824
	CNN	<b>0.850</b>	0.782	0.800
C5 – Need for a Wheelchair	MLP	X	<b>0.840</b>	0.833
	CNN	X	0.827	0.823