

INTRODUCTION

In healthcare, *activities of daily living* (ADLs) describe a set of measurements for evaluating a person's independence and capability in performing various activities related to normal life. It is particularly important when evaluating elderly and disabled.

Computational scientists and clinical practitioners exploit the electronic health records to analyze data and discover patterns. A dominant subset of EHRs data is unstructured clinical notes, which include wealth of information not available in coded form.

OBJECTIVE

This research focuses on finding a method for accurately predicting a particular ADL status of a patient (test case) given a set of progress notes. To do so, models are learned from progress notes representing patients with known values of ADL, as measured by Barthel indices assessed by clinicians.

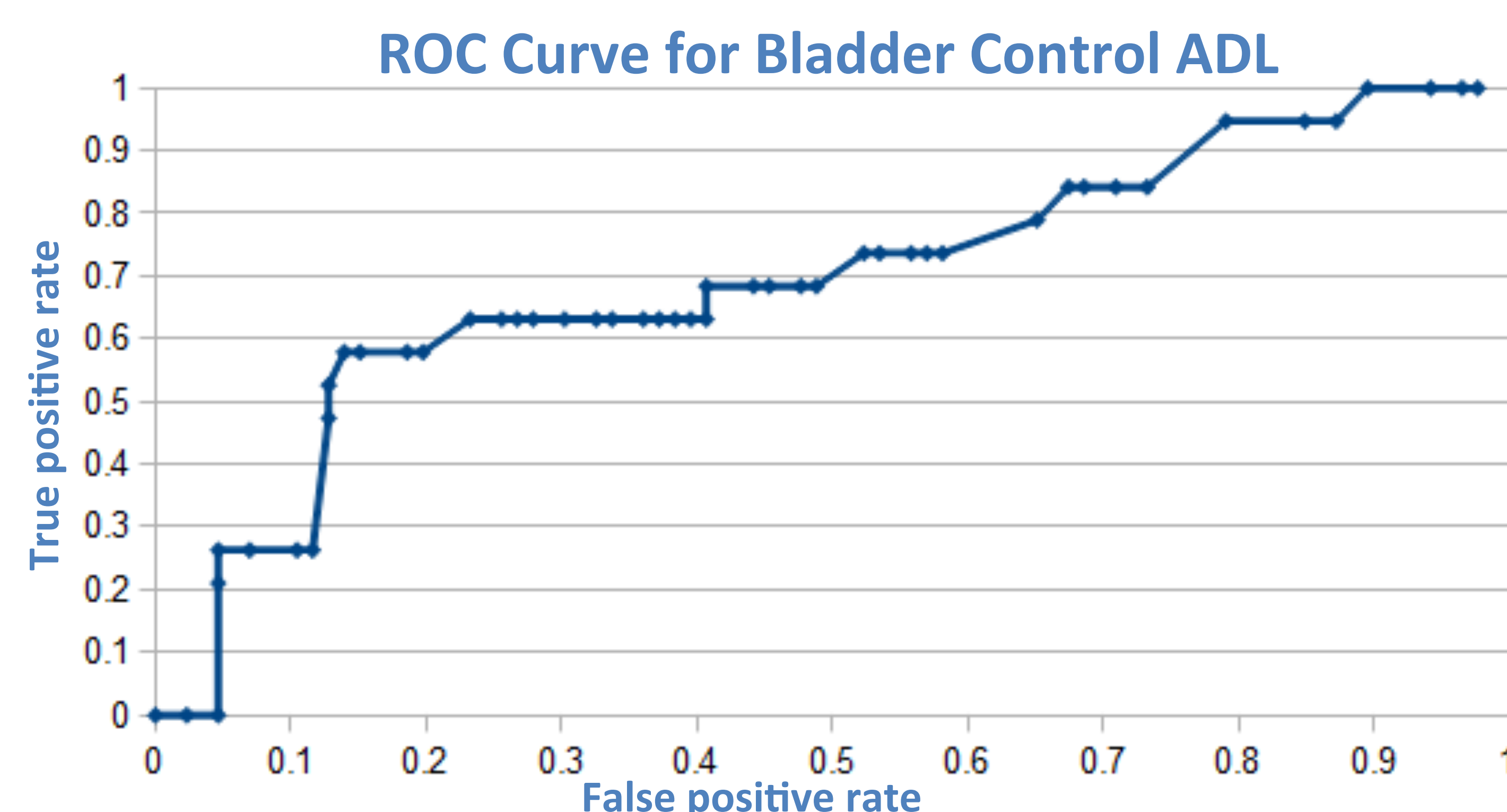
DATA

- The data included 1045 labeled progress notes and 10-fold cross validation technique to test models. Models are learned for each activity from Barthel index separately.
- The dataset is imbalanced: e.g. for Bladder Control ADL, about 776 of the training cases are negative and 164 of them are positive. Remaining 105 notes are used for validation.

Initial Experiments

In the pilot project presented here, N-grams which have high predictive power as measured by likelihood ratios, statistical significances and their expected values have been identified. Design of the system, and obtained classifier's receiver operating characteristic (ROC curve) are presented below.

Most predictive n-grams for Bladder Control ADL	LR
kinesiotherapy progress standard kinesiotherapy ms	14.2
impairment group 2SD0 brain dysfunction	14.2
eud attached bsd draining yellow	9.46
possible interventions va local gi	9.46



Review & Future Work

- The initial work explored several approaches based on likelihood ratios of n-grams. This has led to the observation of limitations related to the size of the dataset used. Also, the results have not been validated for clinical adequacy.
- New features to be incorporated: part-of-speech (POS) from Stanford tagger, section information, word representations: LDA/LSI or Brown clustering, and semantic categories of words based on UMLS lookup, MetaMap or cTAKes outputs.

Sentence
breaking
(Stanford
Core NLP)

Replacement
and removal
filtering with
help of
subject-
matter
experts

Contiguous
word n-gram
generation

Normalization
of numeric
values

Calculation of
the likelihood
ratios of the
n-grams that
appear in the
test case

Deciding on
the class
based on the
informative n-
grams

ADL category	Status
Eating	✓
Transferring	✓
Grooming	✓
Toileting	✓
Bathing	✓
Walking	✓
Dressing	✗
Bowel Incontinence	✓
Bladder Control	✓

