

Patient Condition Modeling in RPM systems: Heart Failure Hospitalization Prediction

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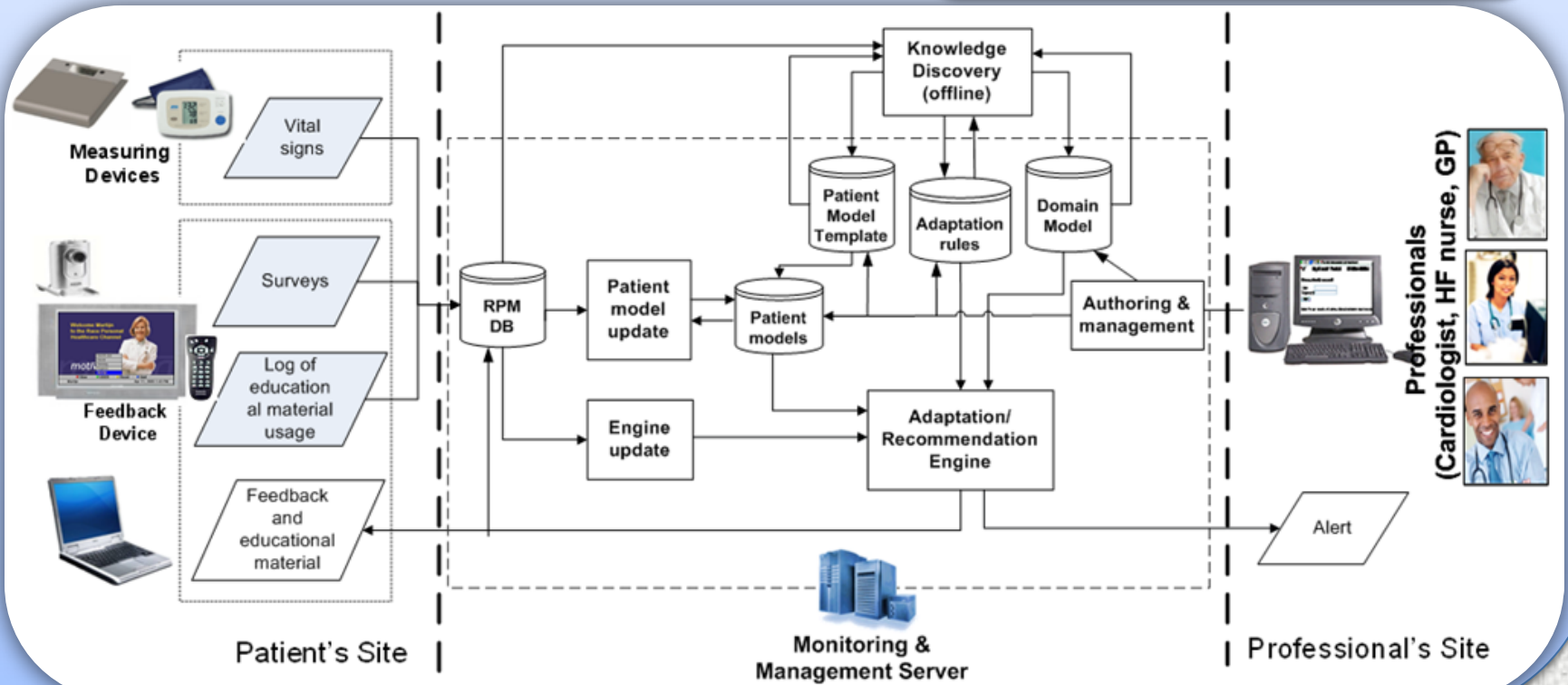
Background and preceding work

Remote patient management (RPM) systems enable

- Monitoring of vital signs of patients at home (blood pressure, weight, etc)
- Providing educational and/or motivational feedback to patients at home
- Alerts to medical professionals of patient's at risk so that they can intervene to prevent worsening of patient's condition and hospitalizations
- All with the aim of improving clinical outcomes of chronic patients, including mortality, hospitalization, and quality of life.

Our recent proposal

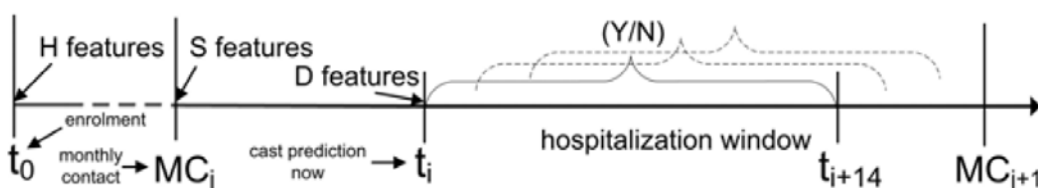
- Possible architecture of next generation personalized RPM systems heavily based on knowledge discovery from RPM data leading to identification of potentially useful features and patterns for patient modeling and adaptation rules



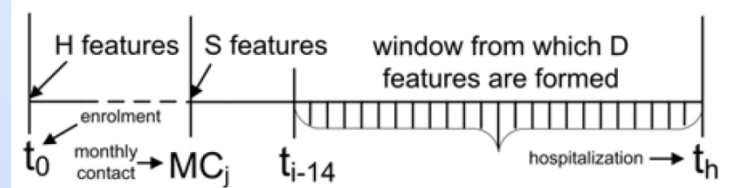
Our approach

Focus of this work is Heart Failure Hospitalization Prediction: On a daily basis, based on the available data about a patient at moment t_i , cast a prediction whether the hospitalization for this patient is likely within next 14 days period (t_{i+1}, t_{i+14}]

Our contribution: We proposed an approach to learn a classifier that utilizes information spread across different data sources and is able outperform the expert-authored rules used to trigger alerts

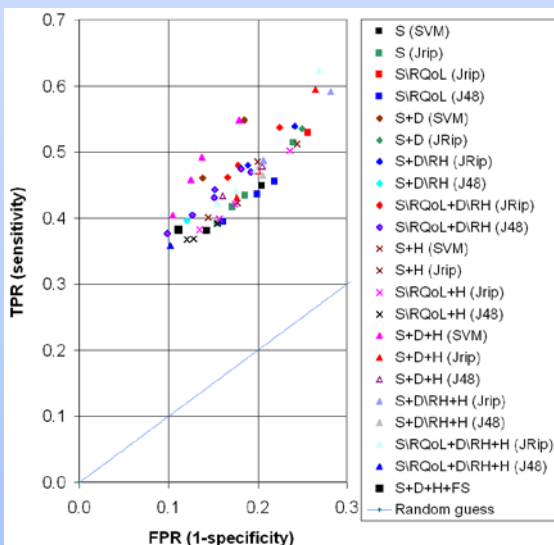


Hospitalization prediction for the following two weeks window

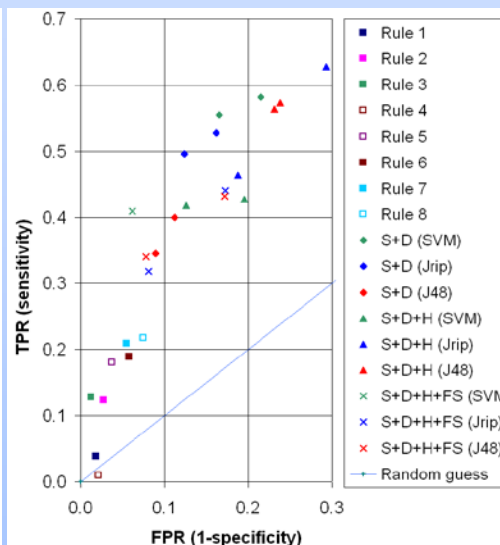


Forming of a positive (HFH took place) training instance

Hospitalization prediction accuracies for different classifiers and feature sets



10CV on training set



test set

Classification model	TPR	FPR	YIndex	HRate	
Rule 1	0.038	0.019	0.019	0.316	
Rule 2	0.123	0.028	0.095	0.211	
Rule 3	0.128	0.012	0.115	0.211	
Rule 4	0.010	0.022	-0.012	0.053	
Rule 5	0.180	0.038	0.143	0.263	
Rule 6	0.189	0.058	0.131	0.316	
Rule 7	0.209	0.055	0.153	0.421	
Rule 8	0.217	0.075	0.142	0.474	
S+D	max TPR	0.582	0.215	0.367	0.526
(SVM)	min FPR	0.555	0.124	0.371	0.474
S+D	max TPR	0.527	0.162	0.365	0.684
(JRip)	min FPR	0.495	0.124	0.371	0.579
S+D	max TPR	0.400	0.112	0.288	0.526
(J48)	min FPR	0.345	0.090	0.256	0.579
S+D+H	max TPR	0.427	0.196	0.232	0.474
(SVM)	min FPR	0.418	0.126	0.292	0.727
S+D+H	max TPR	0.627	0.293	0.334	0.737
(JRip)	min FPR	0.464	0.188	0.276	0.579
S+D+H	max TPR	0.573	0.239	0.334	0.632
(J48)	min FPR	0.564	0.231	0.332	0.632
S+D+H+FS	max TPR	0.432	0.172	0.259	0.632
(SVM)	min FPR	0.409	0.062	0.348	0.579
S+D+H+FS	max TPR	0.441	0.173	0.268	0.632
(JRip)	min FPR	0.318	0.082	0.237	0.368
S+D+H+FS	max TPR	0.432	0.172	0.259	0.632
(J48)	min FPR	0.341	0.078	0.263	0.474

Experimental evaluation on TEN-HMS dataset