

Computational models that predict response to HIV therapy may reduce virological failure and therapy costs in resource-limited settings

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Background

- The long-term effectiveness of antiretroviral therapy (ART) in resource-limited settings (RLS) is challenged by limited access to drugs, diagnostics and expertise
- The adoption of a public health strategy involving simple treatment protocols and the use of clinical or immunological monitoring to detect therapy failure can be associated with deferred/inappropriate treatment switching, accumulation of resistance and increased morbidity and mortality¹⁻⁴
- Computational models have been developed that can predict response to ART from a range of variables including a genotype with circa 80% accuracy^{5,6}
- Models can predict response to ART without a genotype with circa 70-75% accuracy, at least comparable to the accuracy of genotyping with rules-based interpretation⁷⁻⁹
- Models have proved more accurate for patients from familiar settings (with data in the training set) than unfamiliar settings¹⁰

Objectives

- To compare the accuracy of models for patients from 'familiar' settings to those from 'unfamiliar' resource-limited settings (RLS)
- To investigate if the models can identify alternative regimens for cases of failure that have a higher predicted probability of success without additional cost

Methods

- Random forest models used to power the HIV Treatment Response Prediction system (HIV-TRePS)
- Trained with data from 14,891 cases of ART change following virological failure in well-resourced countries
- Input variables: viral load and CD4 count prior to treatment change, treatment history, drugs in the new regimen, time to follow-up and follow-up viral load.
- Output: prediction of the probability of response to ART (<400 copies HIV RNA/ml)

Assessment of models

- Cross-validation
- Independent set of 800 cases from familiar settings

Unfamiliar test sets from RLS

- 231 cases from RLS in Southern Africa
- 375 cases from Romania
- 206 cases from India

Main outcome measure: The area under the ROC curve (AUC)

Modelling alternative regimens

- Baseline data input to models
- Predictions of response obtained for alternative 3-drug regimens comprising only those drugs available in the clinic
- Annual therapy costs used to determine the potential cost effectiveness of this strategy for the Indian cases

Results - 1

Table 1: Accuracy of the models for familiar vs unfamiliar settings

DATA SET	AUC	Overall accuracy	Sensitivity	Specificity
Familiar				
RDI (n=800)	0.77	71%	71%	70%
95%CI	(0.73, 0.80)	(68, 74)	(67, 75)	(64, 75)
Unfamiliar				
Southern Africa (n=231)	0.60**	61%	60%	62%
95%CI	(0.52, 0.69)	(53%, 67%)	(52%, 68%)	(49%, 74%)
Romania (n=375)	0.71	67%	67%	68%
95%CI	(0.66, 0.76)	(62%, 72%)	(60%, 74%)	(60%, 74%)
India (n=206)	0.63*	57%	55%	61%
95%CI	(0.55, 0.71)	(50%, 64%)	(47%, 63%)	(49%, 72%)

*p<0.01 vs RDI 800 **p<0.001 vs RDI 800 (De Long's test)

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Results - 2

Figure 1: ROC curves for the different test sets

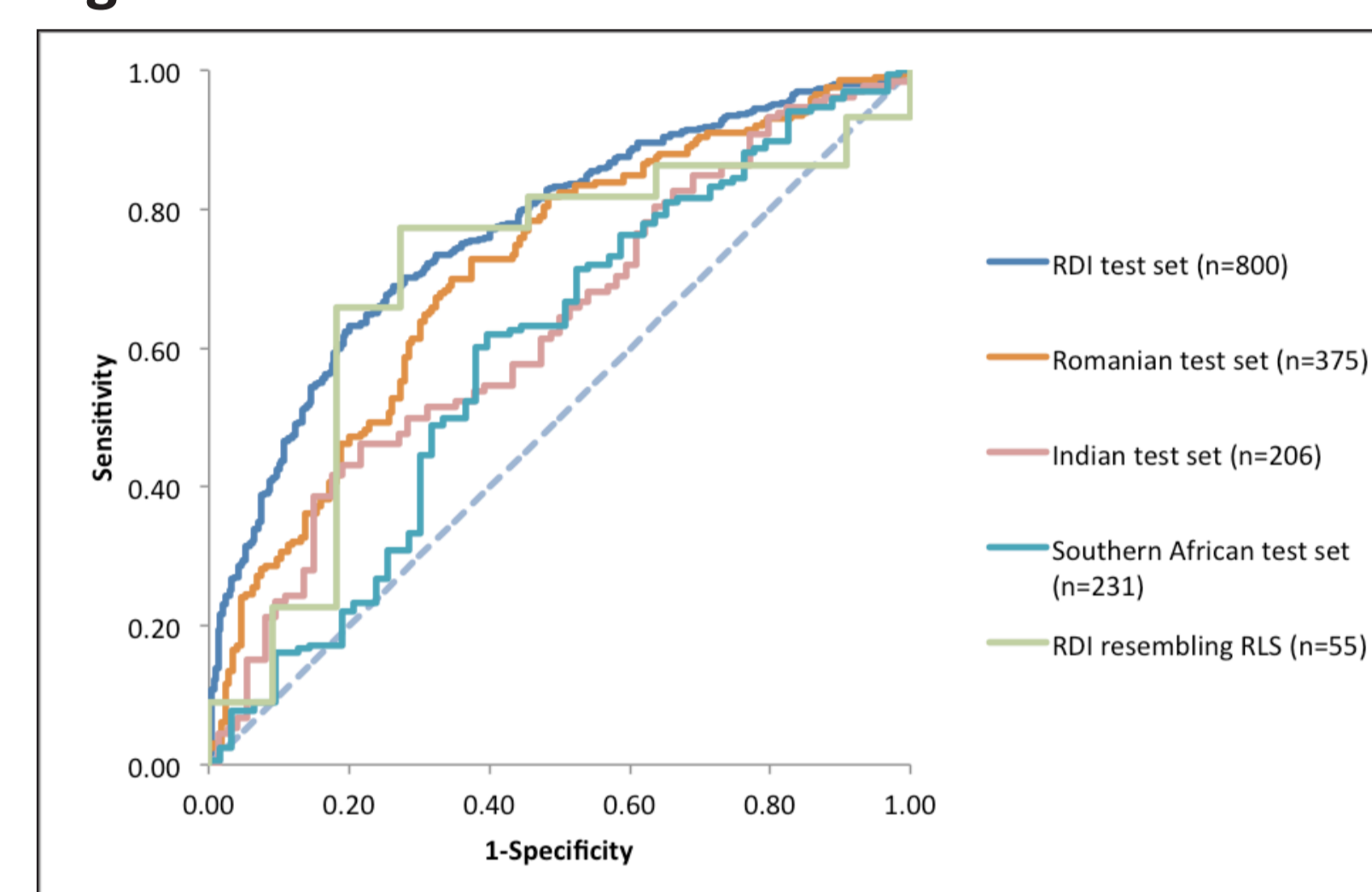


Table 2: Modelling of alternative regimens

	Southern Africa (n=231)	Romania (n=375)	India (n=206)
Number (proportion) of all cases for which the models were able to identify a regimen that was predicted to be effective	217 (94%)	362 (97%)	206 (100%)
No. (%) of cases that failed in the clinic	63 (27%)	176 (47%)	74 (36%)
No. (%) of actual failures for which alternative regimens were found that were predicted to be effective	59 (94%)	164 (93%)	73 (99%)
No. (%) for which alternative regimens were found with higher predictions of response	63 (100%)	176 (100%)	74 (100%)

Table 3: Modelling cost and effectiveness for cases from India

Analysis	All (n=206)	Failures (n=74)
No (%) of cases for which the models were able to identify alternative regimens predicted to be effective	206 (100%)	74 (100%)
No (%) of category 1 alternatives with a higher estimated probability of response than the regimen used in the clinic	175 (85%)	65 (88%)
No (%) of category 2 alternatives where one or more of the alternatives was less costly than the regimen used in the clinic	175 (100%)	65 (100%)
Mean number of alternatives in category 3	10	8
The mean annual cost saving of the least costly regimens in category 3	\$638	\$555

Conclusions

The HIV-TRePS models that predict virological response to ART without a genotype:

1. Showed comparable accuracy to genotyping with rules-based interpretation for patients from unfamiliar RLS
2. Were more accurate for patients from familiar settings suggesting further improvement is possible with more data from RLS
3. Identified alternative regimens that were predicted to be effective for the great majority of cases where the new regimen prescribed in the clinic failed
4. Identified cost-saving alternatives for most cases of failure in India that could fund additional patients' treatment and/or viral load monitoring
5. Have the potential to help optimise antiretroviral therapy in countries with limited resources where genotyping is not generally available.

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 ... and most of all their patients.