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Validation and Refinement of Prediction Models to Estimate Exercise Capacity in Cancer Survivors Using the Steep Ramp Test

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Abstract

Objective: To further test the validity and clinical usefulness of the steep ramp test (SRT) in estimating exercise tolerance in cancer survivors by external validation and extension of previously published prediction models for peak oxygen consumption (VO2peak) and peak power output (Wpeak).

Design: Cross-sectional study.

Setting: Multicenter.

Participants: Cancer survivors (N = 283) in 2 randomized controlled exercise trials.

Interventions: Not applicable.

Main Outcome Measures: Prediction model accuracy was assessed by intraclass correlation coefficients (ICCs) and limits of agreement (LOA). Multiple linear regression was used for model extension. Clinical performance was judged by the percentage of accurate endurance exercise prescriptions.

Results: ICCs of SRT-predicted VO2peak and Wpeak with these values as obtained by the cardiopulmonary exercise test were .61 and .73, respectively, using the previously published prediction models. 95% LOA were ±608mL/min with a bias of 190mL/min for VO2peak and ±59W with a bias of 5W for Wpeak. Modest improvements were obtained by adding body weight and sex to the regression equation for the prediction of VO2peak (ICC, .73; 95% LOA, ±608mL/min) and by adding age, height, and sex for the prediction of Wpeak (ICC, .81; 95% LOA, ±48W). Accuracy of endurance exercise prescription improved from 57% accurate prescriptions to 68% accurate prescriptions with the new prediction model for Wpeak.

Conclusions: Predictions of VO2peak and Wpeak based on the SRT are adequate at the group level, but insufficiently accurate in individual patients. The multivariable prediction model for Wpeak can be used cautiously (eg, supplemented with a Borg score) to aid endurance exercise prescription.

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Many cancer survivors experience severe loss of exercise capacity.\textsuperscript{1,2} This may interfere with performance of daily activities.\textsuperscript{3} Also, poor exercise tolerance is a risk factor for overall mortality.\textsuperscript{4-6} Exercise is recommended for cancer survivors, because it can improve physical fitness and health-related quality of life.\textsuperscript{7}

For optimal exercise prescription for cancer survivors, cancer and cancer treatment–related side effects should be taken into account, as well as basic exercise principles such as specificity, progression, and overload.\textsuperscript{8,9} To evaluate current exercise capacity and to ensure adequate training progression and overload during exercise programs, repeated exercise capacity testing is required.\textsuperscript{9} One of the main indicators of individual exercise capacity is peak oxygen consumption ($VO_{2\text{peak}}$).

The criterion standard of $VO_{2\text{peak}}$ assessment is a maximal incremental exercise test with breath gas analysis and electrocardiography, usually referred to as cardiopulmonary exercise test (CPET).\textsuperscript{10} As the performance of CPET requires specific expertise and specialized equipment, the test is not readily available for exercise prescription in primary care or even in many hospital settings.\textsuperscript{2} As a result, clinicians often turn to less resource-demanding (often submaximal) exercise tests, among which is the steep ramp test (SRT).

The SRT is a short maximal cycle ergometer test that was developed to support interval training prescription for patients with heart failure, without the need for breath gas analysis.\textsuperscript{11} The test uses a fast increasing workload of 25W every 10 seconds. The maximum workload achieved is referred to as the maximal short exercise capacity (MSEC). The largely anaerobic nature of the SRT makes it suitable for dosing interval training and monitoring response.\textsuperscript{11} In a sample of 37 cancer survivors, De Backer et al\textsuperscript{12} showed that the SRT has excellent test-retest reliability and a small SEM. The MSEC also has a strong correlation with $VO_{2\text{peak}}$ and peak power output ($W_{\text{peak}}$) as assessed by a regular CPET,\textsuperscript{12,13} and currently 2 linear regression equations to estimate $VO_{2\text{peak}}$ and $W_{\text{peak}}$ from the MSEC in cancer survivors are available.\textsuperscript{12} The SRT seems to be a promising measure to estimate aerobic exercise capacity and to tailor exercise prescriptions, not only for short bouts of interval training but also for longer endurance training bouts, to the capacity of individual cancer survivors. However, the prediction models as presented by De Backer et al\textsuperscript{12} have not yet been externally validated in a different (and larger) sample of cancer survivors. Also, the predictions of $VO_{2\text{peak}}$ and $W_{\text{peak}}$ by these equations lack precision, with 95% prediction margins of 616mL/min and 63W for $VO_{2\text{peak}}$ and $W_{\text{peak}}$, respectively.\textsuperscript{12} This hampers the use of the SRT for valid estimation of $VO_{2\text{peak}}$ or $W_{\text{peak}}$ and for tailoring other types of aerobic exercises (eg, endurance training with continuous load) in individual patients, because prescribed intensities are often too high or too low. Hence, further improvement of the prediction models is required before the SRT can be considered as a substitute for the CPET.

Better generalizability might be achieved by using a larger sample size to build the regression models. In addition, because age, sex, height, and weight are associated with maximal exercise capacity,\textsuperscript{14} we hypothesized that the accuracy of the prediction of $VO_{2\text{peak}}$ and $W_{\text{peak}}$ based on the MSEC could be improved by including these variables in the regression equation.

Thus, the objectives of this study were to externally validate the previously published prediction models for $VO_{2\text{peak}}$ and $W_{\text{peak}}$\textsuperscript{12} in a larger sample of cancer survivors and to explore whether predictions of $VO_{2\text{peak}}$ and $W_{\text{peak}}$ as well as exercise prescriptions could be improved by extending the regression model.

## Methods

### Patients and measurements

This study used data collected in 2 randomized controlled exercise trials—EXercise Intervention after Stem cell Transplantation study\textsuperscript{15} and Resistance and Endurance exercise After Chemotherapy study\textsuperscript{16,17}—in cancer survivors, which were part of the Alpe d’HuZes Cancer Rehabilitation program.\textsuperscript{18} The EXercise Intervention after Stem cell Transplantation study was a multicenter randomized controlled trial in which 109 patients, who were recently treated with autologous stem cell transplantation for multiple myeloma or (non-)Hodgkin lymphoma, were recruited from 9 hospitals.\textsuperscript{15} The Resistance and Endurance exercise After Chemotherapy study was a multicenter randomized controlled trial in which 277 patients diagnosed with breast, colon, ovarian, cervix, or testis cancer or lymphoma, who had recently completed primary cancer treatment (including chemotherapy), were recruited from 9 hospitals.\textsuperscript{16,17} Approval for these studies was obtained from medical ethics committees of all participating centers, and all participants provided written informed consent.

Both trials included a CPET at baseline and follow-up assessments and used an SRT to support exercise prescription and adaptation. The present analysis uses the first SRT results of each patient, as obtained shortly after the baseline CPET. For the EXercise Intervention after Stem cell Transplantation study, this concerned patients in the intervention group only ($n = 54$). Time between baseline CPET and SRT in all eligible patients ranged between 0 and 41 days. For the purposes of the present study, we only included available data from patients in whom the SRT and CPET were performed within 30 days and who had not yet started exercise training. Consequently, combined tests were available for 283 patients (85%) with a median interval of 8 days (interquartile range, 6–10d).

### Exercise tests

Patients performed both tests on an electronically braked cycle ergometer using standardized protocols. Details of the testing procedures are described elsewhere.\textsuperscript{15,16} In short, the CPET used a ramp protocol with gradually increasing workload that was adjusted to each patient with the aim to achieve the maximum performance within 8 to 12 minutes. Patients were instructed to cycle with a pedal frequency between 60 and 80rpm and were encouraged to continue cycling until exhaustion or inability to maintain the prescribed pedal frequency. During the test, heart rate was monitored continuously using a 12-lead electrocardiogram.
and expired gases were collected breath by breath using gas analysis for O₂, CO₂, and respiration volume. VO₂peak (in milliliters per minute) was determined as the average oxygen uptake during the 15-second interval in which maximum oxygen uptake was attained. Wpeak (in watts) was determined as the highest workload reached during the test.

The SRT started after a 4-minute warming-up process at 10W. During the test, the workload increased by 25W every 10 seconds, starting at 25W. The test ended when the cycling pedal frequency fell below 60rpm. The MSEC was calculated as the workload of the last completed stage plus 2.5W for each second in the current stage.12

Statistical analysis

Descriptive statistics for the sample were calculated using mean and SD, median and interquartile range, or frequency and percentage, as appropriate, on the basis of data type and distribution.

External validation and calibration-in-the-large

We first evaluated the external validity of the prediction models as derived by De Backer et al12 by examining the accuracy of the predictions in the entire sample. Intraclass correlation coefficients (ICCs; 2-way random, absolute agreement, single measure) were calculated between the predicted values (denoted VO₂peak pred and Wpeak pred) and values measured during the CPET (denoted VO₂peak cpet and Wpeak cpet). Second, we updated the intercept of the regression equation to reflect the lower mean VO₂peak and Wpeak in our sample (referred to as calibration-in-the-large).19 Third, we calculated and plotted limits of agreement (LOA) of VO₂peak pred and Wpeak pred with the corresponding values measured during the CPET by using the Bland-Altman method.20 The error correlation (expressing the association between the mean of the values and their difference) was calculated and the corresponding regression line added to the Bland-Altman plot.

Model extension

To extend the 2 regression equations, we conducted multiple linear regression analysis using separate models for VO₂peak pred and Wpeak pred after visually confirming the linear relations between MSEC and VO₂peak pred and Wpeak pred. We started with a full model that included age, height, weight, and sex in addition to MSEC as predictors and used a stepwise backward selection procedure based on the minimization of Akaike’s information criterion (an index of goodness of fit) to select sparse models that best predicted VO₂peak pred and Wpeak pred.19,21 We repeated the LOA analyses as described above using VO₂peak pred and Wpeak pred estimates from the extended multivariable models.

Because estimation of model performance is usually over-optimistic when assessed in the sample in which a model is developed, and regression coefficients are typically too large, we performed internal validation by bootstrapping (100 bootstraps) on the full multivariable model (ie, before stepwise selection). Based on this, uniform shrinkage was applied to the regression coefficients of the reduced model.19 The resulting regression equations for Wpeak pred and VO₂peak pred are presented as a nomogram.

Clinical performance

We evaluated the clinical performance of the original and the newly developed prediction model. For VO₂peak, the percentage of participants whose VO₂peak was misclassified by >1 metabolic equivalent (MET) value (taken as 3.5mL·kg⁻¹·min⁻¹) was calculated. For Wpeak, we evaluated clinical performance on the basis of exercise prescription. A typical exercise prescription at moderate to high intensity would be targeted at 60% of Wpeak. Therefore, we calculated the number of participants for whom an exercise prescription of 60% Wpeak pred fell within ±10 percentage point range of the measured 60% Wpeak.

Sensitivity analyses

We performed 2 sensitivity analyses: The first included only patients with a smaller than 7-day interval between the 2 tests to assess the effect of time between tests on the outcome. The second included patients with breast cancer only (the largest uniform subgroup in the sample) to assess the effect of heterogeneity of cancer diagnoses in our sample. All analyses were performed using the statistical software R version 3.2.1.8

Results

Sample characteristics

Of the 283 participants, 68 (24%) were men. The mean age was 53±11 years, and the majority (162, 57%) were breast cancer survivors (table 1).

External validation

Using the equation for VO₂peak as published by De Backer et al12 (6.7·MSEC+356.7), the ICC of VO₂peak pred with VO₂peak cpet was .61 (95% confidence interval [CI], .41-.74). VO₂peak pred showed a bias of 190mL/min as compared with VO₂peak cpet, which likely reflected the lower mean VO₂peak in our sample than in the derivation sample (21.7mL·kg⁻¹·min⁻¹ vs 29.4mL·kg⁻¹·min⁻¹). After calibration-in-the-large, the 5% LOA were ±705mL/min. There was a negative error correlation: r = -.22 (fig 1A).

The ICC of Wpeak pred with Wpeak cpet was .73 (95% CI, .67-.78) using the equation for Wpeak as published by De Backer et al (0.65·MSEC−3.88). The Bland-Altman analysis indicated a small bias of 5W, despite a larger difference in mean Wpeak between our sample and the sample in which the equation was derived (141W vs 183W). After calibration-in-the-large, 95% LOA were ±59W, and there was a negative error correlation: r = -.14 (fig 1B).

Model extension

The final multivariable model for the prediction of VO₂peak included weight and sex in addition to the MSEC (table 2). The explained variance (R²) of this model was .58, and the ICC of VO₂peak pred with VO₂peak cpet was .73 (95% CI, .67-.78). Using this model for prediction, LOA improved to ±608mL/min, but the negative error correlation remained: r = -.39 (see fig 1A). After shrinkage, the equation was as follows: 676.8+3.92·MSEC+5.02·weight−327.6·female (using 1 for “female” and 0 for “male”). The equation is presented as a nomogram (fig 2A).
For $W_{\text{peak}}$, the final model included height, sex, and age in addition to the MSEC (see table 2). The $R^2$ of this model was .67. The ICC of $W_{\text{peak}}$ pred with $W_{\text{peak}}$ cpct was .81 (95% CI, .76-.84). Compared to the original equation, LOA improved to ±48W, but the error correlation was larger: $r = -.33$ (see fig 1B). After shrinkage of the coefficients, the equation was $0.33 \times \text{MSEC} + 124 \times \text{height (m)} - 22.4 \times \text{female} - 0.47 \times \text{age} - 107$. This equation is also presented as a nomogram (fig 2B).

### Clinical performance

Using the original equation, 56% of the patients were misclassified by $>1$MET. An exercise prescription aimed at achieving $60\% W_{\text{peak}}$ was inaccurate by $>10$ percentage points (eg, $<50\%$ or $>70\%$ of $W_{\text{peak}}$) in 43% of the sample. After calibration-in-the-large, this reduced to 44% and 39% of the patients, respectively. Using the reestimated, multivariable prediction models, the percentage further reduced to 36% and 32%, respectively.

### Sensitivity analyses

No significant and meaningful differences in the LOA or clinical applicability were observed when restricting the sample to patients with a measurement interval $<7$ days ($n = 109$) or breast cancer survivors only ($n = 139$).

### Discussion

This study examined the validity and clinical applicability of regression-based estimates of VO$_{2\text{peak}}$ and $W_{\text{peak}}$ in cancer survivors by using the SRT. In particular, we evaluated the external validity of previously derived prediction models for this population and aimed to improve predictions by extending the regression models.

Our results confirm the linear relations of MSEC with VO$_{2\text{peak}}$ and $W_{\text{peak}}$ that were also described in previous studies of cancer survivors, patients with type 2 diabetes mellitus, and healthy children and adolescents. As far as we know, no validation studies have been conducted to externally validate the prediction rules derived in these studies.

The bias observed in the LOA analysis shows that the accuracy of the predicted values VO$_{2\text{peak}}$ pred and $W_{\text{peak}}$ pred affected if the mean values of VO$_{2\text{peak}}$ cpct and $W_{\text{peak}}$ cpct in the target population are different from those of the population in which the equation was derived (miscalibration in the large). Although predictions improved by extending the regression models, they are still suboptimal. The agreement between the values as predicted by the extended models and the actual values could be considered sufficiently high at the group level. However, higher agreement (ICCs, $>0.9$) is usually considered preferable for decision making at the individual level. Also, the LOA, which have a stronger influence on clinical usability than the ICCs, are still wide. This is illustrated by the high percentage of patients that were misclassified by $>1$MET in the present study.

### Table 1 Descriptive statistics of the sample ($N = 283$)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex: male</td>
<td>68 (24)</td>
</tr>
<tr>
<td>Age (y)</td>
<td>53±11</td>
</tr>
<tr>
<td>Height (cm)</td>
<td>170±9</td>
</tr>
<tr>
<td>Body weight (kg)</td>
<td>77.6±14.6</td>
</tr>
<tr>
<td>Cancer type</td>
<td></td>
</tr>
<tr>
<td>Breast cancer</td>
<td>162 (57.2)</td>
</tr>
<tr>
<td>Colon cancer</td>
<td>49 (17.3)</td>
</tr>
<tr>
<td>Lymphomas</td>
<td>56 (19.8)</td>
</tr>
<tr>
<td>Ovarian cancer</td>
<td>8 (2.8)</td>
</tr>
<tr>
<td>Cervix cancer</td>
<td>4 (1.4)</td>
</tr>
<tr>
<td>Testis cancer</td>
<td>4 (1.4)</td>
</tr>
<tr>
<td>Treatment</td>
<td></td>
</tr>
<tr>
<td>Chemotherapy</td>
<td>283 (100)</td>
</tr>
<tr>
<td>Radiotherapy</td>
<td>123 (43)</td>
</tr>
<tr>
<td>Surgery</td>
<td>227 (80)</td>
</tr>
<tr>
<td>Stem cell transplantation</td>
<td>32 (11)</td>
</tr>
<tr>
<td>Immunotherapy</td>
<td>51 (18)</td>
</tr>
<tr>
<td>Hormone therapy</td>
<td>114 (40)</td>
</tr>
<tr>
<td>Weeks since treatment</td>
<td>8±3.7</td>
</tr>
<tr>
<td>VO$_{2\text{peak}}$ (mL/min)</td>
<td>1713.7±476.0</td>
</tr>
<tr>
<td>VO$_{2\text{peak}}$ (mL·kg$^{-1}$·min$^{-1}$)</td>
<td>21.7±5.7</td>
</tr>
<tr>
<td>$W_{\text{peak}}$ (W)</td>
<td>141±43.1</td>
</tr>
<tr>
<td>MSEC (W)</td>
<td>231±60.4</td>
</tr>
</tbody>
</table>

NOTE. Values are mean ± SD or n (%).

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![Fig 1](https://example.com/fig1.png)  
**Fig 1** Bland-Altman plots for the estimation of VO$_{2\text{peak}}$ (A) and $W_{\text{peak}}$ (B) after reestimation and extension of the prediction model. Dashed lines indicate 95% LOA; dotted lines indicate 95% confidence interval of the limits, and the diagonal line indicates linear regression fit of difference vs mean.
Table 2  Regression coefficients for multivariable linear regression model predicting \( V_{O2peak} \) and \( W_{peak} \) using the SRT

<table>
<thead>
<tr>
<th>Variable</th>
<th>( \beta )</th>
<th>95% Confidence Interval for ( \beta )</th>
<th>( \beta ) After Shrinkage</th>
</tr>
</thead>
<tbody>
<tr>
<td>( V_{O2peak} )</td>
<td>MSEC</td>
<td>3.97</td>
<td>3.29 to 4.65</td>
</tr>
<tr>
<td>MCFS</td>
<td>Body weight</td>
<td>5.09</td>
<td>2.37 to 7.80</td>
</tr>
<tr>
<td>Sex: female</td>
<td>(-332)</td>
<td>(-432) to (-232)</td>
<td>(-328)</td>
</tr>
<tr>
<td>( W_{peak} )</td>
<td>MSEC</td>
<td>0.34</td>
<td>0.28 to 0.40</td>
</tr>
<tr>
<td>MCFS</td>
<td>Age</td>
<td>(-0.47)</td>
<td>(-0.75) to (-0.19)</td>
</tr>
<tr>
<td>MCFS</td>
<td>Height</td>
<td>126</td>
<td>82.1 to 169</td>
</tr>
<tr>
<td>MCFS</td>
<td>Sex: female</td>
<td>(-22.6)</td>
<td>(-31.9) to (-13.3)</td>
</tr>
</tbody>
</table>

In addition to quantification of peak exercise capacity, identification of the limiting system (cardiovascular, pulmonary, musculoskeletal, psychological) and assessment of exercise-induced cardiovascular risks can be of relevance in the context of cancer rehabilitation. Such information cannot be obtained using the SRT. Also, without measuring gas exchange, it can be difficult to ascertain peak performance. Therefore, despite its advantage of being quick and easily applicable, the SRT should not be considered an alternative to a CPET when a valid measurement of \( V_{O2peak} \) is needed for an individual patient or when the exercise capacity limiting system needs to be identified for clinical decision making.

As expected, the clinical applicability of the original prediction model for the estimation of \( W_{peak} \) in cancer survivors was also insufficient for accurate aerobic exercise prescription (based on a constant percentage of estimated \( W_{peak} \)) in our sample. This could be improved considerably by calibration-in-the-large and even constant percentage of estimated \( W_{peak} \) in our sample. This could be implemented more readily by calibration-in-the-large and even further after extending the equation with height, age, and sex. Nevertheless, clinicians should be aware of the possibility of over- or underestimation of exercise capacity when using \( W_{peak} \) for individual exercise prescription. The observed negative error correlation implies an overestimation of \( V_{O2peak} \) and \( W_{peak} \) in cancer survivors with lower cardiorespiratory fitness and underestimation in those with high cardiorespiratory fitness. The presented nomogram (see fig 2) is expected to have sufficient accuracy in about two thirds of cancer survivors.

One could argue on the basis of our results that CPET remains a preferred choice for exercise prescription. However, several studies have shown that even with available CPET data, accurate exercise prescription for cancer survivors, for example, based on the percentage of maximum heart rate, can be challenging because of changes in exercise response.24-26 Also, repeated testing is required to ensure sufficient overload during the course of the exercise program, and repeated CPETs are difficult to implement in regular care. In a recent study, repeated SRTs was used in combination with Borg scores of perceived exertion to prescribe fixed load endurance exercise during chemotherapy. Patients after this program successfully maintained physical fitness throughout chemotherapy.27 Clearly, more studies are needed to examine the accuracy of different ways to tailor exercise prescription, and the role of different exercise tests, in cancer survivors.

### Study strengths

Strengths of this study include the relatively large sample that allowed for a robust analysis strategy, including sensitivity analyses by test interval and tumor type, and the use of a heterogeneous sample of cancer survivors. Although it is likely that the performance of the extended prediction models will be somewhat less when applied to other patients, our results are expected to translate reasonably to clinical practice because shrinkage was applied to the regression coefficients.19,21

### Study limitations

This study also had some limitations that should be noted. We used data that were gathered in the context of randomized controlled trials evaluating the effects of exercise interventions in cancer survivors.15,17 The timing of the tests relative to each other was not fixed, but depended on trial logistics. We tried to limit the potential effect of time between measurements by selecting only those cancer survivors who were tested on the CPET and SRT within a maximum range of 30 days. Because the sensitivity analysis in survivors with a maximum time interval between the tests of 7 days showed comparable results, we believe that the

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Fig 2  Nomograms for the prediction of \( W_{peak} \) (A) and \( V_{O2peak} \) (B). The value of each variable corresponds to a number of points on the top scale. Points are summed. The total points can then be used to read the linear predictor (estimated \( W_{peak} \) or \( V_{O2peak} \)) from the bottom scale. For example, for a patient who achieves an MSEC of 250W (56 points), weighs 86kg (13 points), and is a female (0 points), the total points are 69, which corresponds to a linear predictor of \(-1760\, \text{mL/min.}\)
timing had limited effect on our findings. Second, although the exercise tests were carried out according to standardized protocols, the SRT measurements were conducted by different physical therapists offering the exercise programs. This may have introduced random error in the measurements, reducing the observed validity of the SRT, but it does resemble the way the SRT will be used in clinical practice.

The sample was limited to cancer survivors who had recently completed treatment. It is unclear whether the results are generalizable to cancer survivors during active treatment. Factors such as fatigue, anxiety, nausea, or lack of motivation may be more prominent during active treatment, and may affect the workload achieved on the SRT more profoundly than that on the CPET, because of the short duration and steep increments used in the SRT. It is conceivable that this will influence the relation of the MSEC with VO2peak and Wpeak. Future research should further explore these issues.

For cancer survivors who completed treatment, the presented nomogram provides added value to support personalized endurance exercise intensity prescription based on the SRT. However, the nomogram-predicted values should be considered a starting point and further adaptation (eg, using Borg scores) may still be necessary for individual patients.

Conclusions
The results of this study suggest that although predictions of VO2peak and Wpeak based on the SRT could be considered acceptable at the group level, they do not provide sufficiently accurate estimations of endurance exercise capacity in individual cancer survivors. Nevertheless, the prediction of Wpeak using the presented multivariable prediction rule or the associated nomogram can be used cautiously (eg, supplemented with a Borg score) to aid endurance exercise prescription in this population.

Supplier
a. R version 3.2.1; R Foundation for Statistical Computing.

Keywords
Exercise test; Exercise therapy; Neoplasms; Rehabilitation; Validation studies

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