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## Defining Accelerometer Thresholds for Activity Intensities in Adolescent Girls

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### Abstract

**Purpose**—To derive a regression equation that estimates metabolic equivalent (MET) from accelerometer counts, and to define thresholds of accelerometer counts that can be used to delineate sedentary, light, moderate, and vigorous activity in adolescent girls.

**Methods**—Seventy-four healthy 8th grade girls, age 13-14 yr, were recruited from urban areas of Baltimore, MD, Minneapolis/St. Paul, MN, and Columbia, SC, to participate in the study. Accelerometer and oxygen consumption ( $\dot{V}O_2$ ) data for 10 activities that varied in intensity from sedentary (e.g., TV watching) to vigorous (e.g., running) were collected. While performing these activities, the girls wore two accelerometers, a heart rate monitor and a Cosmed K4b2 portable metabolic unit for measurement of  $\dot{V}O_2$ . A random-coefficients model was used to estimate the relationship between accelerometer counts and  $\dot{V}O_2$ . Activity thresholds were defined by minimizing the false positive and false negative classifications.

**Results**—The activities provided a wide range in  $\dot{V}O_2$  (3-36 mL·kg<sup>-1</sup>·min<sup>-1</sup>) with a correspondingly wide range in accelerometer counts (1-3928 counts·30 s<sup>-1</sup>). The regression line for MET score versus counts was MET = 2.01 + 0.00171 (counts·30 s<sup>-1</sup>) (mixed model R<sup>2</sup> = 0.84, SEE = 1.36). A threshold of 1500 counts·30 s<sup>-1</sup> defined the lower end of the moderate intensity (~4.6 METs) range of physical activity. That cutpoint distinguished between slow and brisk walking, and gave the lowest number of false positive and false negative classifications. The threshold ranges for sedentary, light, moderate, and vigorous physical activity were found to be 0-50, 51-1499, 1500 -2600, and >2600 counts·30 s<sup>-1</sup>, respectively.

**Conclusion**—The developed equation and these activity thresholds can be used for prediction of MET score from accelerometer counts and participation in various intensities of physical activity in adolescent girls.

## Keywords

ACTIVITY MONITORS; ENERGY EXPENDITURE; PHYSICAL ACTIVITY ASSESSMENT; FITNESS; OXYGEN CONSUMPTION

There is considerable public health interest in the physical activity levels of children and adolescents in the United States, in part because there is an increased prevalence of obesity (16) and because obesity has been associated with lower physical activity (6). Research has documented low physical activity levels of adolescents (2), particularly in girls (8). Furthermore, a precipitous decline in physical activity levels in adolescent girls was recently reported with almost no physical activity reported by ages 18 and 19 yr (9). These concerns have sparked interest in interventions to improve physical activity in this age and gender group. The recently initiated Trial of Activity for Adolescent Girls (TAAG) is a multi-site school- and community-based intervention designed to prevent the decline in physical activity in adolescent girls. This paper reports on a preliminary study using girls from the catchment areas of three TAAG field centers to define accelerometer thresholds for physical activity.

Moderate-to-vigorous intensity physical activity (MVPA) levels result in positive health changes. For example, adolescent participation in activity has been shown to reduce the rate of increase in obesity (13), lower total cholesterol (13), lower circulating insulin levels (20), and slow the age-related increase in blood pressure (7). However, our present state of knowledge precludes good field methods for differentiating between intensity levels. The use of accelerometers or activity monitors has become a promising method to accomplish this goal. The Actigraph (Manufacturing Technologies Inc. Health Systems) accelerometer, previously called the CSA accelerometer, records activity counts in a given time interval, for example, half-minute or minute intervals. Calibrating an accelerometer against a “gold standard” measure of physical activity intensity, such as oxygen consumption ( $\dot{V}O_2$ ), requires a controlled experiment by which  $\dot{V}O_2$  and accelerometer counts are observed simultaneously over a range of activity intensities. A calibration curve relating  $\dot{V}O_2$  to accelerometer counts can be fit to these data, and used subsequently for the estimation of metabolic equivalent (MET) score from accelerometer counts.

Laboratory-based studies have compared accelerometers to energy expenditure during physical activity in adults (5,15,26,27) and children (17,22,23). However, studies using  $\dot{V}O_2$  as the criterion measure to establish cut-points of physical activity specifically in children are few (17). In 26 boys and girls 7-15 yr of age, Puyau et al. (17) reported an equation to predict energy expenditure from accelerometer counts, as well as intensity thresholds for two accelerometers (Actigraph and Actiwatch). Moderate activity ranged from 3200 to 8200 counts·min<sup>-1</sup> for the Actigraph (17). No study has involved a large representative sample of middle-school girls.

In the present study, accelerometer counts were calibrated against measured energy expenditure ( $\dot{V}O_2$ ) during a variety of activities. A regression equation to estimate MET (metabolic equivalent) values from accelerometer counts was created. The purpose of the study was to then generate the cut-points defining sedentary, light, moderate, and vigorous activity using accelerometer counts.

## METHODS

### Participants

In the spring of 2001, 74 healthy 8th grade girls, ages 13-14 yr, were recruited from urban areas of Baltimore, MD, Minneapolis/St. Paul, MN, and Columbia, SC. A sample was recruited that was diverse in terms of height, weight, race, and history of sport participation. None of the

participants had physical limitations. The protocol was approved by each participating institution's Human Subjects Review Board. Consent to participate was obtained from one parent, and assent was obtained from each participant.

### Instrumentation

Age and ethnicity were obtained from the participant by questionnaire. Following a standardized protocol, the participant's weight was measured twice to the nearest 0.1 kg on an electronic scale (Seca, Model 770, Hamburg, Germany). Height was measured twice to the nearest 0.1 cm using a portable stadiometer (Shorr Height Measuring Board, Olney, MD). The average of the two measures was used.

Oxygen consumption ( $\dot{V}O_2$ ) and heart rate were measured using a portable, breath-by-breath, metabolic unit available from Cosmed (Model K4b2, Rome, Italy). The K4b2 is a lightweight (925 g) system that can be worn on the back. The unit includes a POLAR Pacer heart rate transmitter (Pt. Washington, NY). The calibration of the Cosmed followed standard procedures across all sites. The Cosmed unit was calibrated with standard gases before each visit and recalibrated with gases for the biking outside (an activity described below). The system has been previously validated (11).

All Actigraph (Manufacturing Technologies Inc. Health Systems, Model 7164, Shalimar, FL) accelerometers were checked at the Coordinating Center for similarity of basic functional condition using a standard laboratory shaker before being sent to each field site for subsequent use in this study. The test consisted of shaking the accelerometers secured on a mechanical shaker for 15 min at a low speed (similar to a walk) and 15 min at a high speed (similar to a slow jog). The variation between all units was less than 10%, which is within the limits of acceptability by the manufacturer. Each participant wore two accelerometers, one over each hip. Analysis of the data comparing the right versus the left hip yielded very similar results. All pairwise correlations between predicted METs from accelerometry using counts from the right hip, left hip, and the average of the two were 1.0. Therefore, all accelerometer values used for the analysis were the mean of right and left hip accelerometer counts. The monitors were attached to a belt worn around the waist. For each visit, the monitors were initialized before the first activity, and data were uploaded from the monitor to a PC after performance of all activities was complete. Activity counts were stored in 30-s time intervals. The 30-s epoch was chosen because children tend to do activities in short bursts and more sporadic than adults.

### Study design

Participants attended two visits to obtain accelerometer and  $\dot{V}O_2$  data for all 10 activities (Table 1). The selected activities were chosen to provide varied intensity from sedentary to vigorous and to reflect typical activities performed by middle school girls. Each participant completed both visits within a 2-wk time frame. Participants were required to fast for 3 h before each visit. At the first visit, participants were given an overview of the physical activity assessments, followed by height and weight measurement. The participant was then fitted with two Actigraph monitors and the Cosmed system and given an opportunity to become familiar with breathing through the mask. Once the prescribed activities and measurements began, accelerometer counts, heart rate, and  $\dot{V}O_2$  were recorded simultaneously on a 30-s basis during the entire 7 min of each activity. Rest periods between activities were allowed as necessary to ensure that heart rate was below 100bpm before beginning the next activity.

### Activities

Table 1 provides a brief description of the activities assessed. Resting  $\dot{V}O_2$  was collected for 15 min in a dimly lighted, quiet room, with the participant in a reclined position. Participants were not allowed to sleep and several minutes of rest were provided before metabolic

measurement occurred. Minutes 3-15 were then used for data analysis. Activities were completed in ascending intensities, starting with rest and ending with running. Sites were allowed to vary which activities took place during each visit, as long as the intensity of the activity was ascending. The protocol for visit 1 generally included assessments of rest, watching television, playing computer games, a slow walk, and a stair walk at a specified pace (see Table 1). Activities completed during the second visit generally included sweeping the floor, a brisk walk, step aerobics, shooting baskets with a basketball, and running. The biking activity took place at the end of either the first or second visit. For two activities (watching television and playing the computer game), the participant was seated in a chair. All other activities were in the upright standing position (except bicycling) and were set at specific paces. For all activities other than rest,  $\dot{V}O_2$ , heart rate and accelerometer counts were collected for 7 min, with the mean value of the last 4-7 min used for data analysis. Participants were considered to be at steady-state energy expenditure during this time (28). In order to standardize the activity (pace and movement), when-ever possible a staff member completed the activity along with the participant. The walking/running speeds for the participants were defined using paces set between marked distances with verbal guidance used to cue the participant to speed up or slow down as needed. For bicycling, a speedometer was used to monitor pace. Data collection was monitored continuously throughout all activities by trained staff. The staff recorded the exact time of the measurements, as well as marking the event button on the Cosmed unit.

The Physical Work Capacity test ( $PWC_{170}$ ) involved a standardized protocol (18) on a cycle ergometer (Monark, Model 818E, Varberg, Sweden). The test measures fitness, so that we could include in our sample participants within a wide range of fitness levels. This was generally completed on the first visit. The participants completed at least two 3-min stages during which  $\dot{V}O_2$  was measured continuously and recorded every 30 s. Cadence was set to a metronome at 60 cycles·min<sup>-1</sup>. The initial cycling workload was based upon the participant's weight and subsequent workloads were determined by the heart rate at the end of the stage according to the established  $PWC_{170}$  protocol. Heart rate was recorded during the last 10 s of each minute for each 3-min stage. The test was terminated when the heart rate reached 165 bpm or greater. From the results of the PWC, maximal aerobic power ( $\dot{V}O_{2max}$ ) was predicted using the methodology outlined by McMurray et al. (12) and Mocellin et al. (14). This information was also used to interpret the intensities of each activity (% $\dot{V}O_{2max}$ ) for classification of low, moderate, and high intensity activity.

### Statistical analysis

Because resting  $\dot{V}O_2$  of youth is higher than adults (21), MET values were first computed based on the girl's actual resting  $\dot{V}O_2$  rather than the standard adult MET (3.5 mL·kg<sup>-1</sup>·min<sup>-1</sup>). These MET values were then used for further analyses. A random-coefficients model was used to characterize the relationship between accelerometer counts and  $\dot{V}O_2$  over a range of activity intensities, while accounting for the dependence among repeated measurements taken on the same girl (10,19). Letting  $y_{ij}$  denote the response for the  $j$ th observation on the  $i$ th participant, the model can be written as

$$y_{ij} = \alpha + \beta x_{ij} + a_i + b_i x_{ij} + e_{ij} \quad [1]$$

where  $\alpha$  and  $\beta$  are fixed effects that denote the population average intercept and slope, and  $a_i$  and  $b_i$  represent each participant's random intercept and slope deviations about the corresponding population parameters. Restricted maximum likelihood estimates of random effects and maximum likelihood estimates of fixed effects were obtained using the MIXED procedure in SAS Version 8.2 (SAS, Cary, NC). Coefficients associated with quadratic and cubic trends ( $\gamma$ ,  $\lambda$ ) in third-order polynomial model,  $y_{ij} = \alpha + \beta x_{ij} + \gamma x_{ij}^2 + \lambda x_{ij}^3 + a_i + b_i x_{ij} + e_{ij}$ , were not found to be significantly different from zero. Similarly, age, body mass index, and

race were not found to contribute significantly to the fit of the model. Thus, only results from the unadjusted linear mixed model are reported here.

A range of Actigraph counts were considered as candidate cut-points for defining intensity levels. The method of selecting an optimal cut-point was based on two criteria: 1) correct identification of the target or higher intensity activities (i.e., sensitivity); and 2) correct exclusion of lower intensity activities (i.e., specificity).

Three approaches were compared for their potential in choosing the lower threshold for moderate intensity physical activity. Approach 1 defined low intensity activities as watching TV, playing a computer game, sweeping the floor, and slow walking, and moderate or higher intensity activities as brisk walking, stair walking, shooting baskets, step aerobics, and running. Counts observed for low intensity activities that exceeded a given threshold were classified as false positives, and counts observed for moderate or higher intensity activities that fell below the same threshold were classified as false negatives. Thus, a false positive was an observation that did not represent moderate activity but was interpreted as moderate activity, and a false negative was an observation known to represent moderate activity that was interpreted as less than moderate activity. Thresholds between 1100 and 1600 counts·30 s<sup>-1</sup> were examined. Bicycling was excluded from the analysis due to the poor correlation between accelerometer counts and  $\dot{V}O_2$  during bicycling. Approach 2 was identical to approach 1, but step aerobics was also excluded from the analysis, also due to the poor correlation between accelerometer counts and  $\dot{V}O_2$ . Approach 3 focused only on slow and brisk walk as indicators of low and moderate intensity activities, respectively, based upon the literature defining brisk walking as a moderate activity (24). A bootstrap validation procedure (3) was used to assess the degree to which the cut-points would be expected to apply to a new sample of adolescent girls. One hundred bootstrap samples, each half the size as the original sample, were generated by drawing observations at random and with replacement from the original sample. The rate of misclassification (false positive and false negatives) was determined for all bootstrap random samples, and 5th and 95th percentile is reported.

The approaches were evaluated by examining the number of activity observations classified as false positive and false negative. The optimal threshold was the cut-point that both balanced and minimized the number of false positives and false negatives.

Based upon our experience in determining the moderate activity threshold using the Actigraph accelerometer, the cut-points for the other intensity levels were established in the same manner. To choose the upper threshold for light activity, counts observed for playing computer games (i.e., sedentary activity <1.5 METs) that exceeded a given threshold were classified as false positives, and counts observed for slow walking (i.e., light activity 3 METs) that fell below the threshold were classified as false negatives. To define the lower threshold of vigorous activity, counts observed for brisk walking (i.e., moderate activity 3.8 METs) above a given threshold were classified as false positives, and counts for running (i.e., vigorous activity 6 METs) below the threshold were classified as false negatives. As with the moderate threshold, the optimal thresholds for the other intensities were the cut-points that both balanced and minimized the number of false positives and false negatives.

## RESULTS

### Participant characteristics

The participant characteristics are provided in Table 2. Thirty-three Caucasians, 27 African-Americans, 4 Asians, and 10 multi-racial girls participated in the study. There was a wide range in BMI (14-40 kg·m<sup>-2</sup>) by design. As well, the predicted  $\dot{V}O_{2max}$  values from the PWC-170



test reflect a very wide range in fitness, from low fitness ( $18.5 \text{ mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$ ) to high fitness ( $65.8 \text{ mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$ ) levels.

### Accelerometer counts and MET scores

Table 3 includes the  $\dot{V}\text{O}_2$ , MET score and accelerometer counts, expressed both in half minutes ( $\text{counts}\cdot 30 \text{ s}^{-1}$ ) and minutes ( $\text{counts}\cdot\text{min}^{-1}$ ) for each activity. All accelerometer counts are presented as the mean of the right and left hip accelerometer counts. The activities are shown in increasing value of  $\dot{V}\text{O}_2$  and MET scores, with the lowest values for watching TV and computer games, and the highest values for walking the stairs and running. The chosen physical activities provided a wide range in MET scores (1-7.8) with a correspondingly wide range in accelerometer counts (1-3928  $\text{counts}\cdot 30 \text{ s}^{-1}$  or 1-7867  $\text{counts}\cdot\text{min}^{-1}$ ). The accelerometer values for bicycling (369  $\text{counts}\cdot 30 \text{ s}^{-1}$ ) and step aerobics (1371  $\text{counts}\cdot 30 \text{ s}^{-1}$ ) were considerably lower than what would be expected from the  $\dot{V}\text{O}_2$  values. For each activity, the counts per minute were not exactly twice the counts per 30 seconds because two sequential 30-s periods do not always have identical counts.

Figure 1 shows the regression lines for MET score versus accelerometer counts for each girl (thin lines) and the over-all regression line (heavy line). Figure 2 illustrates the regression line and 95% CI for the same analysis. Formula (1) estimates MET score using counts. As noted earlier, the results were similar when the data were analyzed separately for the left and right hip accelerometers. Using the left side,  $\text{MET} = 1.98 + 0.00175 (\text{counts}\cdot 30 \text{ s}^{-1})$ , and using the right side,  $\text{MET} = 2.10 + 0.00165 (\text{counts}\cdot 30 \text{ s}^{-1})$ .

$$\begin{aligned} \text{MET} &= 2.01 + 0.00171 (\text{counts} \cdot 30 \text{ s}^{-1}) \\ \text{MET} &= 2.01 + 0.000856 (\text{counts} \cdot 60 \text{ s}^{-1}) \end{aligned} \quad [2]$$

The mixed model concordance correlation coefficient (25), corresponding to an  $R^2$  in standard linear regression, was 0.84 (SEE = 1.36).

### Threshold for moderate intensity physical activity

Examination of thresholds from 1100 to 1600  $\text{counts}\cdot 30 \text{ s}^{-1}$  for approach 1 (excluding bicycling) gave high false positives and high false negatives (data not shown). By excluding step aerobics and bicycling (approach 2), the number of false positives and false negatives was lower than by excluding bicycling alone (Table 4); even so, approach 2 still gave high false positives (3-43) and high false negatives (6-32). Approach 3, which distinguished between slow and brisk walking, gave the lowest and most similar number of false positives and false negatives (Table 5). Under Approach 3, the number of false positives was similar to the number of false negatives for the 1470-1580  $\text{counts}\cdot 30 \text{ s}^{-1}$  accelerometer range (four to six false positives and three to eight false negatives). At the mid-point of this range (i.e., 1530 counts), there were five false positives and four false negatives. At a count of 1500, six false positives and three false negatives and the lowest total number of classification errors occurred. Therefore, a threshold of 1500  $\text{counts}\cdot 30 \text{ s}^{-1}$  was accepted as the lower end of the moderate intensity range of physical activity, using approach 3.

### Thresholds for sedentary, light, and vigorous intensity physical activity

The summary of threshold counts for sedentary, light, moderate, and vigorous physical activity for the accelerometer is presented in Table 6. Thresholds for data collection periods of both 30-s intervals and 1-min intervals are shown. A lower threshold for light activity of 50  $\text{counts}\cdot 30 \text{ s}^{-1}$  resulted in no false positives and false negatives. By default, counts below 50 were defined as sedentary. The optimal lower threshold for vigorous physical activity (and the upper limit moderate activity) was observed at a threshold of 2600  $\text{counts}\cdot 30 \text{ s}^{-1}$ , with only eight brisk



walking observations above the threshold (false positives) and nine running observations below the threshold (false negatives).

## DISCUSSION

In this study, we developed a regression equation to predict MET score from the accelerometer counts. The model fit was good, as evidenced by a mixed model equivalent to  $R^2$  for linear regression of 0.84, even so, the large standard error of  $\pm 1.36$  METs might suggest that the equation be used only as a gross estimate of energy expenditure. The use of the MET score for the prediction is a unique approach as previous papers typically use caloric expenditure in kilocalories (5,17). The use of MET values also allows for the use of the Compendium of Physical Activity to estimate energy expenditure over a large range of activities (1). Our equation will be useful to others interested in predicting a MET score with their own accelerometer data. It has already been useful in defining moderate-to-vigorous physical activity for the main trial data from TAAG. As well, this equation can be used in reducing the accelerometer data collected in TAAG.

There is relatively little other research using accelerometry to predict energy expenditure in children. In one study (17), activity energy expenditure was predicted from Actigraph accelerometer counts using controlled laboratory conditions in a room calorimeter as well as free-living field activities. Both boys and girls of varying ages participated and a high correlation was found ( $R^2_{\text{adj}} = 0.75$ ). In adults, other laboratory-based studies have compared accelerometers to energy expenditure during physical activity (5,15,26,27). Freedson et al. (5) predicted energy expenditure ( $\text{kcal}\cdot\text{min}^{-1}$ ) measured in adults while walking at different speeds on the treadmill from Actigraph counts and BMI, reporting an  $R^2$  of 0.82. In two studies of adults by Welk et al. (26,27), validation coefficients between several accelerometers (Actigraph, Biotrainer, and Actitrac) with energy expenditure measured under treadmill and field conditions were also moderate to high (range  $r = 0.62$ - $0.91$ ). Thus, our equation to estimate MET levels from accelerometer counts in adolescent girls appears comparable to previous reports in children and adults.

In order to define our activity thresholds appropriately, we utilized the  $\dot{V}O_2$  values above each girl's own resting metabolic rate. One MET is defined as  $3.5 \text{ mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$  for adults (1), and in our study, this value was slightly higher ( $3.8 \pm 0.9 \text{ mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$ ). Utilizing the adult value for 1 MET might have biased the estimated energy cost of the physical activities measured in these adolescent girls. Thus, we chose to use the participants' actual resting values to estimate MET levels.

Our three different approaches for determining our thresholds identified activities that were falsely classified. In approach 1 where bicycling was excluded, counts for watching TV, playing a computer game, sweeping the floor, and slow walking that were above the threshold were classified as false positive for moderate intensity activity, and counts for brisk walking, stair walking, shooting baskets, and running below the same threshold were classified as false negative. There were high numbers of false negatives, mostly related to step aerobics. This may have been expected since the accelerometer senses a walk (or step), but the participant had to lift her weight up a 6-inch step each step; thus, metabolic rate would be increased above walking. In approach 2, step aerobics and bicycling were both removed from the analysis. The numbers of false positives and false negatives were lower than excluding bicycling alone. This suggests that the accelerometer does not respond appropriately during activities in which body motion is somewhat independent of exercise intensity. Approach 3 had the lowest number of false positive and false negative classifications. This approach used differences between a slow and brisk walk to identify thresholds. Brisk walking has previously been categorized as a moderate activity (24). In particular, the recent guidelines for meeting the minimum level of

physical activity levels use brisk walking as an example of a moderate activity for individuals (2,4). Thus, although we tested many activities with a wide range of intensities, a potentially limiting factor in the study is that we relied on differentiating between two walking speeds to determine our moderate threshold.

We experienced some difficulties using the accelerometer to establish the thresholds for light and vigorous activity. In our attempt to define a vigorous threshold, we designed the study to include activities that would be at an intensity > 6 METs. Previous literature has defined 6 METs as vigorous (24). As shown in Table 3, shooting baskets and the stair walking yielded MET values greater than 6 METs, but the activity counts were very low. As adolescents perform activities in variable ways, it is evident that the accelerometer does not work well in evaluating the intensity of such activities. Perhaps accelerometers attached to the waist lack in their ability to account for upper extremity work. It is also possible that the accelerometers might detect movement when in fact the adolescent was sedentary (e.g., car/bus travel). This could have influenced the number of false negatives for the sedentary threshold. Thus, accelerometers are limited in their ability to estimate activity levels or metabolic rates during cycling, step aerobics, stair stepping, or other similar activities where the accelerometer counts are less accurate.

One study (17) evaluated similar cut-points in children using the Actigraph accelerometer during treadmill walking/running and jogging outside at their own speed. A slightly higher value for the lower threshold for moderate activity was reported (<3200 counts·min<sup>-1</sup>). This corresponded to <0.05 kcal·kg<sup>-1</sup>·min<sup>-1</sup> (17). Their reported threshold for vigorous activity was >8200 counts, with sedentary defined as <800 counts·min<sup>-1</sup>. Our values were lower and may be due to the types and intensities of activities chosen to determine these thresholds. For example, vigorous activity included treadmill walking (that specifically controlled speed and step rate) and jogging on a track (17), whereas we did not use a treadmill. Greater variability would be expected with our run at 5 mph (8 kph). Our participants had to follow a set pace and were verbally encouraged to either speed up or slow down if the pace deviated.

Because we evaluated the fitness levels of the participants using a cycle ergometer test (PWC<sub>170</sub>), we were able to verify how our chosen thresholds relate to  $\dot{V}_{O_{2max}}$ . Moderate physical activity should correspond to 40-60% of  $\dot{V}_{O_{2max}}$ . In these participants with a predicted average  $\dot{V}_{O_{2max}}$  of ~39 mL·kg<sup>-1</sup>·min<sup>-1</sup>, 40-60% would be equivalent to 16 -23 mL·kg<sup>-1</sup>·min<sup>-1</sup>. The measured  $\dot{V}_{O_2}$  values for the slow walk in our study are below this range whereas the value for brisk walking (4.3 METs and  $\dot{V}_{O_2} = 15.6$  mL·kg<sup>-1</sup>·min<sup>-1</sup>) was at the low end for the range of 40 - 60%  $\dot{V}_{O_{2max}}$ . Thus using slow and brisk walking as activities to separate light from moderate activity to define our activity thresholds appears to be appropriate. We can also compare our chosen activities and measured  $\dot{V}_{O_2}$  to a percentage of this estimated  $\dot{V}_{O_{2max}}$ . The 11 chosen activities range from 11% to 79% of predicted  $\dot{V}_{O_{2max}}$ . Thus, we can be confident that our prediction equation for MET score was developed from the entire range of exercise intensities. One limitation of our prediction is that these values were predicted from cycle ergometry, which is lower than what would be predicted from treadmill walking. Nevertheless, we did include a wide range of exercise intensities into the design of the study.

One unique component of this study was the setting of the accelerometers to collect activity counts in 30-s intervals. Children's activities tend to be spontaneous, sporadic, and done in shorter intervals. Our rationale for this shorter data collection interval was that sensitivity might be increased and the activities might be better detected with the 30 s rather than 1 min collection time. We presented the data also in counts per minute for our thresholds, as this has been the standard in the literature (5,17).

Based on our study, a prediction equation for estimating MET score from accelerometer counts is now available for use in adolescent girls. The accelerometer thresholds to define girls'

physical activity levels can also be used to determine the amount of sedentary, light, moderate, and vigorous physical activity. This type of determination is useful for surveillance research or to determine the effects of physical activity interventions designed to promote activity.

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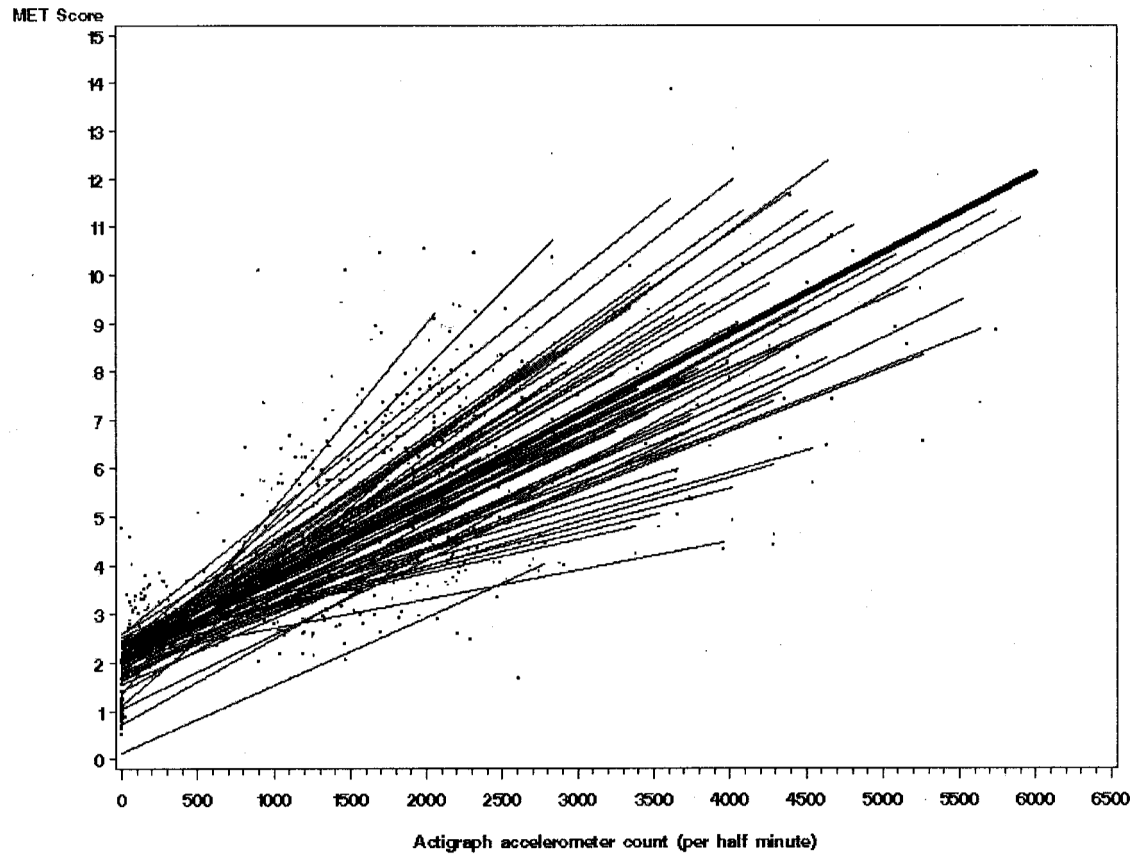
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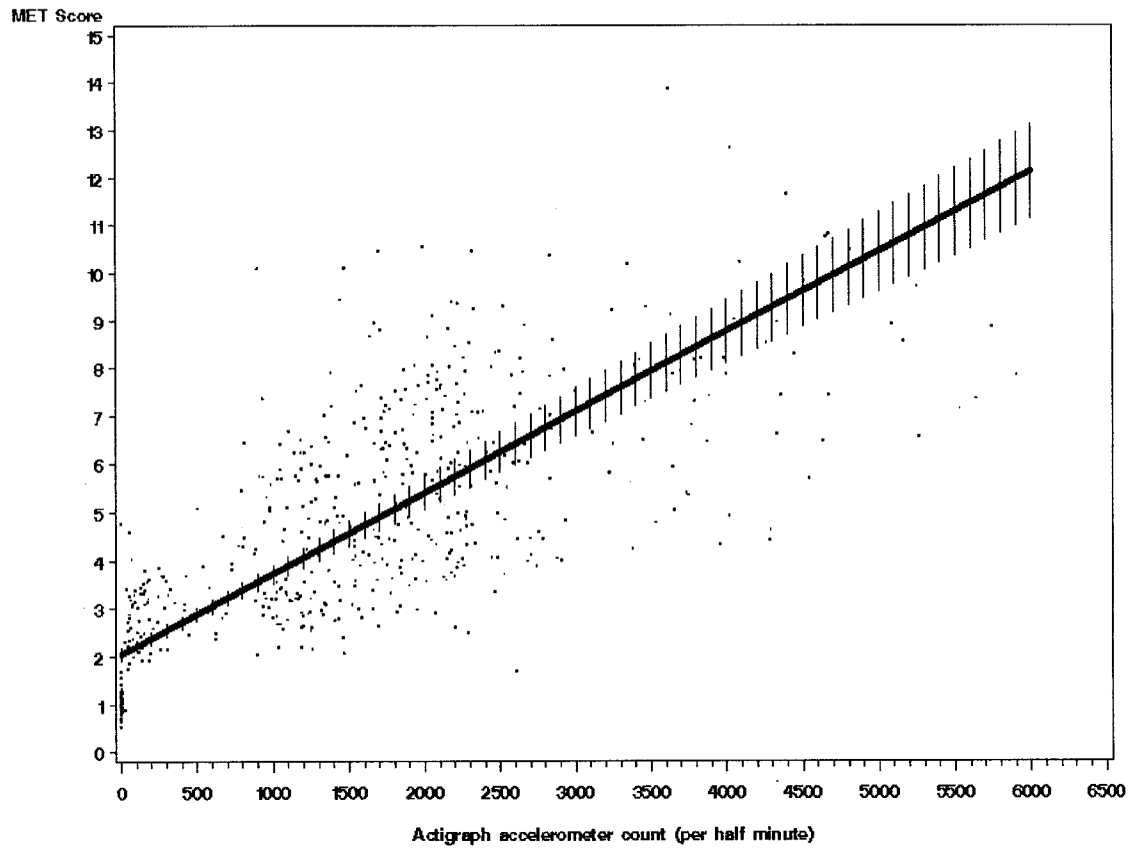
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**FIGURE 1.** Regression line for MET score versus accelerometer (Actigraph) counts (random regression line for each girl and overall regression line).



**FIGURE 2.**  
Regression line and 95% CI for MET score versus accelerometer counts.

TABLE 1

## Study description of activities

Activity	Description
Rest	Lie on bed, still, awake, quiet environment
Watch TV	Sit in chair, watch movie video (e.g., comedy)
Play computer game	Sit in chair, play Chip's Challenge game on PC
Sweep floor	Sweep confetti on floor continuously using broom to a specified location and repeating
Slow walk	Walk at 2.5 mph., participant's pace kept close to that speed with the guidance of timed distances
Brisk walk	Walk at 3.5 mph., participant's pace kept close to that speed with the guidance of timed distances
Step aerobics	Step aerobics video, both arm and leg movements, 1 step bench, first practice and instruction from staff before actual measurement, all exercised to the same 7 min of video
Bicycling	Ride bicycle outside at 12 mph, cyclometer on bike used to monitor speed, no steep hills, relatively flat
Shoot baskets	Shooting the basketball and retrieving the basketball continuously, using girls' size basketballs
Stair walk	Climb stairs in a staircase at 80 steps·min <sup>-1</sup> to music tape, without holding the rail except for balance
Run	Run at 5.0 mph on track or marked area, paced by staff with the guidance of timed distances



**TABLE 2**Subject characteristics ( $N = 74$ )

	Mean (SD)	Range
Age (yr)	14.1 (0.3)	13.4-15.4
Weight (kg)	60 (15.5)	35-102
Height (cm)	160 (6.6)	146-177
BMI ( $\text{kg}\cdot\text{m}^{-2}$ )	23.4 (5.8)	14.4-40.0
$\text{VO}_{2\text{max}}$ predicted ( $\text{mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$ )	38.5(9.9)	18.5-65.8

TABLE 3

$\dot{V}O_2$ , MET score, and accelerometer counts for each activity

Activity	N	$\dot{V}O_2$ (mL·kg <sup>-1</sup> ·min <sup>-1</sup> )		% of $\dot{V}O_{2max}$	MET Score*		Accelerometer Count·30 s <sup>-1</sup>		Accelerometer Count·min <sup>-1</sup>	
		Mean (SD)	Mean (SD)		Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)		
1. Rest	64	3.8 (0.9)	0 (0.7)							
2. Watch TV	67	3.9 (0.9)	1.0 (0.1)	11	1.0 (0.1)		1 (1.7)		2 (4.6)	
3. Play computer game	71	4.2 (2.0)	1.1 (0.5)	11	1.1 (0.5)		1 (3.2)		2 (6.4)	
4. Sweep floor	72	11.0 (2.2)	3.0 (0.6)	30	3.0 (0.6)		203 (173)		400 (339)	
5. Slow walk	72	11.8 (2.5)	3.2 (0.8)	32	3.2 (0.8)		1180 (280)		2356 (537)	
6. Brisk walk	73	15.6 (2.3)	4.3 (1.0)	43	4.3 (1.0)		2085 (379)		4160 (748)	
7. Step aerobics	71	21.2 (3.8)	5.7 (1.4)	57	5.7 (1.4)		1371 (457)		2750 (916)	
8. Bicycling	66	21.3 (7.6)	5.9 (2.3)	59	5.9 (2.3)		369 (297)		741 (594)	
9. Shoot baskets	69	24.3 (4.8)	6.5 (1.6)	67	6.5 (1.6)		2002 (584)		4005 (211)	
10. Stair walk	72	24.6 (3.7)	6.6 (1.4)	68	6.6 (1.4)		2138 (393)		4271 (788)	
11. Run	65	29.5 (5.0)	7.8 (2.2)	79	7.8 (2.2)		3928 (170)		7867 (375)	

All accelerometer (Actigraph) values are the mean of right and left hip accelerometer counts; counts per 30 seconds are not necessarily half of the counts per minute because consecutive 30-s periods do not always have identical counts.

\* MET score = (Activity  $\dot{V}O_2$  mL·kg<sup>-1</sup>·min<sup>-1</sup>)/(Individual Resting  $\dot{V}O_2$  mL·kg<sup>-1</sup>·min<sup>-1</sup>).

Choosing the threshold for moderate-to-vigorous intensity physical activity by examining the false positives and false negatives based for approach 2 (eight activities)

TABLE 4

Accelerometer (counts·30 s <sup>-1</sup> )	True Positive (N)	True Negative (N)	False Positive (N)*	False Negative (N)**	Error Rate (%)	Validation Error Rate (%)*** (5th, 95th Percentile)
1100	239	273	43	6	8.7	6.4, 11.4
1150	242	273	40	6	8.2	5.9, 10.3
1200	252	270	30	9	7.0	4.6, 8.9
1250	259	270	23	9	5.7	3.5, 7.5
1300	263	269	19	10	5.2	3.4, 7.0
1350	268	267	14	12	4.6	2.9, 6.7
1400	269	266	13	13	4.6	3.0, 6.7
1450	274	261	8	18	4.6	3.0, 6.6
1500	276	258	6	21	4.8	2.8, 6.5
1550	277	250	5	29	6.1	4.1, 7.9
1600	279	247	3	32	6.2	3.9, 8.2

\* Counts measured under the protocols for watching TV, playing a computer game, sweeping the floor, and slow walking that were above the threshold for moderate intensity activity were considered as false positives.

\*\* Counts measured under the protocols for brisk walk, stair walk, shoot baskets, and run that were below the threshold for moderate intensity activity were considered as false negative (exclude step aerobics and bicycling).

\*\*\* Based on 100 bootstrap samples, each half the size as the original sample.

Choosing the threshold for moderate-to-vigorous intensity physical activity by examining the false positives and false negatives based for approach 3 (slow and brisk walking)

TABLE 5

Accelerometer (counts·30 s <sup>-1</sup> )	True Positive (N)	True Negative (N)	False Positive (N)*	False Negative (N)**	Error Rate (%)	Validation Error Rate (%)*** (5th, 95th Percentile)
1100	29	73	43	0	29.7	21.7,37.4
1150	32	73	40	0	27.6	19.8,34.9
1200	42	72	30	1	21.4	14.2,28.2
1250	49	72	23	1	16.6	9.6,22.0
1300	53	72	19	1	13.8	7.9,19.8
1350	58	72	14	1	10.3	6.0,16.7
1400	59	72	13	1	9.7	5.6,13.9
1450	64	71	8	2	6.9	3.5,11.5
1500	66	70	6	3	6.2	2.5, 9.4
1550	67	65	5	8	9.0	4.0,13.8
1600	69	65	3	8	7.6	2.7,12.5

\* Counts measured under the protocol for slow walk that were above the threshold for moderate intensity activity were considered as false positive.

\*\* Counts measured under the protocol for brisk walk that were below the threshold for moderate intensity activity were considered as false negative.

\*\*\* Based on 100 bootstrap samples, each half the size as the original sample.

**TABLE 6**

Summary of intensity threshold counts for sedentary, light, moderate, and vigorous physical activity (PA) for the Actigraph accelerometer

Activity Level	Accelerometer (counts·30 s <sup>-1</sup> )	Accelerometer (counts·min <sup>-1</sup> )
Sedentary PA	<50	<100
Light PA	51-1499	101-2999
Moderate PA	1500-2600	3000-5200
Vigorous PA	>2600	>5200