Modeling human-bed interaction: the predictive value of anthropometric models in choosing the correct bed support

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Abstract. The sleep system (i.e. the combination of mattress and bed base) is an important factor of the sleep environment since it allows physical recuperation during sleep by providing proper body support. However, various factors influence the interaction between the human body and the sleep system. Contributing factors include body dimensions, distribution of body weight and stiffness of the sleep system across the mattress surface. During the past decade, the rise of several new bedding technologies has made it increasingly difficult for the consumer to select a proper sleep system. Therefore, this study presents a method to model human-bed interaction in order to objectively predict the ideal sleep system for a particular individual. The proposed method combines a personalized anthropometric model with standardized load-deflection characteristics of mattress and bed base. Results for lateral sleep positions show a root mean square deviation of 11.9 ± 6.1 mm between modeled spine shapes and validation shapes, derived from 3D surface scans of the back surface. The method showed to be a reliable tool to individually identify the sleep system providing superior support from a variety of possible mattress–bed base combinations.

Keywords: ergonomic design criteria, sleep system optimization, personalized body support, mechanical bed characteristics, bed support

1. Introduction

Over the past ten years, an increased effort has been taken to incorporate ergonomic aspects in the design of sleep systems (i.e. the combination of mattress and bed base). Whereas mattresses with a homogeneous stiffness distribution used to be the standard, nowadays most systems in Western countries consist of distinct comfort zones in order to properly contour the human body [7]. In addition, the importance of a personalized approach when purchasing a sleep system is being emphasized more and more (e.g. different body types need different sleep systems) [11, 25]. However, various technical innovations – with new materials being deployed in all components of the sleep system – have hampered a well-considered purchase of the consumer. Consequently, there is a strong demand for scientifically validated decision criteria to assign a sleep system to a specific person.

The most important function of sleep systems is to provide proper support in a way that allows the human body to recover from daily activities. Optimal recovery is achieved when the spine is in its natural, unloaded shape, allowing muscle relaxation and rehydration of the intervertebral discs [16, 14]. In this context, optimizing body support refers to aligning the spine towards its shape during stance, yet with a somewhat flattened lumbar lordosis due to the changed working axis of gravity [4]. Since spinal alignment is the result of the complex biomechanical interaction of the human body and the sleep system, it is influenced by a variety of factors, including not only the mechanical characteristics of the bedding system (more specifically the stiffness distribution),
but also the subject’s weight distribution and body dimensions. Ideally, all these factors are considered when purchasing a new sleep system.

Several methods have been described in literature to assess bed comfort, such as spine shape reconstruction [13, 3, 10], pressure mapping [21, 22, 15], electromyography (EMG) [13, 12] and subjective evaluations [17, 1]. In this study, spinal alignment will be considered as the main criterion to assess bed comfort for the following reasons. First, spinal alignment provides a global measure of the deformation of the human body while sleeping [18]. Second, although pressure distribution is of primary concern for the prevention of pressure ulcers in bedridden patients [21], it is sufficient to avoid concentrated pressure peaks when considering a general (healthy) population. Third, at present the validity of EMG to assess body support on sleep systems is rather poorly documented and remains subject to discussion [13].

This study presents a method to model human-bed interaction in order to objectively predict the ideal mattress–bed base combination for a particular person. The method combines a personalized anthropometric model of the sleeper with information on the mechanical behavior of mattress and bed base. Since it is based on standardized mechanical characteristics (load-deflection properties), it is independent of bedding technology. Consequently, the presented method enables to individually identify the sleep system with superior support properties from a range of distinct mattress–bed base combinations.

2. Methodology

2.1. Participants

Eighteen subjects (nine males, nine females; aged 28.5 ± 4.7 years) participated in the study. Anthropometric information was determined by means of silhouette extraction in both the sagittal and frontal plane (Ikélo, Custom8, Leuven, Belgium).

2.2. Bedding systems

Three types of bed bases were included: 1) a homogeneous box-spring, 2) a multi-zone slatted base, and 3) a multi-zone mesh base. They were combined with three standard types of mattresses: 1) a multi-zone pocket spring mattress, 2) a multi-zone latex mattress, and 3) a homogeneous polyurethane foam mattress. Hence, a total of nine mattress–bed base combinations were accounted for.

2.3. Simulation

2.3.1. Human body shape model

A generic surface model of the human body was used that consists of consecutive superellipses representing the transverse cross sections of the human body. Personalization is achieved by optimizing the semi-diameters and ellipse order to match the measured body contours (Figure 1). Prior research [24] fully described this procedure and validated the resulting models with 3-D surface scans of the trunk, showing a mean unsigned distance of 9.77 mm between modeled and scanned surface meshes.

The surface model is further extended by integrating a distribution of body weight over the different body parts. Appropriate slice masses were determined based on density values of human body segments described in literature [6]. The volume-based slice masses were scaled in such way that their sum corresponded to the measured body weight – reduced with the estimated weight of head and feet, which were not modeled.

2.3.2. Characterization of sleep system components

Load-deflection curves were determined for each bed base and each mattress separately using a flat and circular stamp (diameter 210 mm, rounded with a 5 mm radius at the lower edge). Travel speed of the
stamp was kept constant at 120 mm/min. Five discrete locations were characterized along the center line of the sleep systems at respectively 45 cm, 65 cm, 75 cm, 90 cm and 100 cm of the head end. Load-relief cycles were repeated at least three times and approximated by polylines. Only the third cycle was retained for further calculations. Figure 2 illustrates the measurement set-up and provides some examples of load-deflection characteristics for three locations on a box-spring base and a pocket spring mattress.

2.3.3. Simulation of spine shape

In order to simulate human-bed interaction based on the personalized human model and the load-deflection characteristics several considerations are taken into account.

First of all, the combined mechanical behavior of mattress and bed base is simulated based on the separate load deflection characteristics. On that account, the local stiffness ratio of mattress and bed base has shown to be a predicting factor to determine the equivalent force that is exerted by the mattress on the bed base when the complete sleep system is loaded by a specific force.

A second consideration involves the distinct contact surface (both area and shape) of the human body segments with respect to the circular stamp that was used to determine the load-deflection characteristics. Therefore two correction factors were defined as follows:

\[
C_1 = \frac{A_{\text{eff}}}{A_{\text{ref}}} \\
C_2 = \frac{a}{h}
\]  

(1)

with \(A_{\text{eff}}\) the contact area of the indentation stamp, \(A_{\text{ref}}\) the effective contact area of the body segment, \(a\) the lateral semi-diameter of the superellipse and \(h\) the distance between the effective contact line \((T_{\text{eff}})\) and the center of the superellipse. Figure 3 clarifies these factors. The enclosing rectangle of the superellipse defines the effective contact line of the cross section with the mattress surface \((T_{\text{eff}})\). Numerical integration over subsequent superelliptic cross sections determines the effective contact area \(A_{\text{eff}}\). Both correction factors were applied to the weights of the corresponding body segments.

The human model was segmented at 20, 30, 45 and 55 cm below the shoulder joint to assure that the centers of the body segments coincided with the locations for which indentation characteristics were determined (with the shoulder joint coinciding with the location at 45 cm of the head end). Each segment consisted of five consecutive superellipses corres-
ponding to a segment length of 10 cm. Based on the corrected weights of the body segments their displacement was calculated. Finally, the line through the centers of the body segments represented the simulated shape of the spine.

2.4. Validation

2.4.1. Measurements

Validation of the predicted spine shape was provided by a 3-D surface scan of the back surface (zSnapper, Vialux, Chemnitz, Germany). Based on surface curvature information the locations of the following landmarks were determined: the vertebra prominens, the dimples of the posterior superior iliac spine and the sacrum point [5]. The line through the spinous processes was reconstructed by means of an active contour model iterating on a weighted combination of surface curvature information and lateral asymmetry information [8]. Prior research validated this approach with full spine CT measurements, reporting a root mean square deviation of 2.6 mm in lateral sleep positions [9].

In order to compare the validation shapes with the simulated spine shapes, they were subsampled to discrete points corresponding to the locations of the body segments at 20 cm, 30 cm and 45 cm below the shoulder point. Figure 4 demonstrates the simulation and validation of a particular mattress–bed base combination for one of the participants.

2.4.2. Analysis

An ergonomic bed score for lateral sleep postures (EBS_L) was defined based on the weighted combination of the angle of the least square line through the spine points (α) and the angle between the lumbar and the thoracic part of the spine (γ):

$$EBS_L = 10 \cdot \left[ 1 - \exp \left( -\frac{0.4}{5 \cdot (|\alpha| + 0.25 \cdot |\gamma|)} \right) \right]$$

This score (between 0 and 10) accounts for both the orientation of the spine with respect to the horizontal plane and for the deviation of the spine from a straight line. Based on the ESB_L, a ranking of the tested mattress–bed base combinations was determined for each person separately.

3. Results

Table 1 gives an overview of the anthropometric data that was gathered to personalize the models along with the calculated weights of the different parts of the trunk. Figure 5 shows the average simulated and measured spine shapes along with the corresponding variation whereas table 2 presents the descriptive statistics of the ergonomic bed scores derived from the predicted and validated spine shapes. Both figure 5 and table 2 illustrate the validity of the presented method to predict quality of bed support in general. A more detailed, quantitative comparison of predicted spine shapes with the validation shapes revealed a root mean square deviation of 11.9 ± 6.1 mm. In addition, it can be inferred from figure 6 that no effect of bed base type nor mattress type was noted on the accuracy of spine shape prediction (p = 0.22).

Finally, the rankings based on the measured shapes showed an overall correspondence of 85 % with the rankings based on the simulated shapes. More specifically, the method correctly identified for
16 out of 18 subjects the sleep system offering superior support. The two cases in which the best combination was not predicted, showed a very small difference in ESB_L with the second best combination (0.29 and 0.25 respectively), indicating that the two superior combinations provided similar support.

4. Discussion

This study shows that it is feasible to individually assess quality of support provided by a particular mattress–bed base combination. During the past decade the importance of a personalized approach when choosing a sleep system has increasingly been emphasized. However, no clear decision criteria have been proposed to objectively quantify the effect of different sleep system components (e.g. mattress and bed base) on quality of body support. One of the underlying reasons for this lack of objective criteria is the existence of various influencing factors that determine the complex interaction between the human body and the sleep system. Consequently, this study aimed at providing a method to individually determine spine shape when lying in a lateral position on a particular sleep system. Its main asset is that it is independent on the type of mattress or bed base, making it suitable for decision support across brand specific technologies.

The presented method is based on the working hypothesis that body support is determined by individual body features on the one hand (e.g. body contours, weight distribution,…) and the mechanical behavior of the sleep system on the other hand. The former is accounted for by means of a personalized human model [24], whereas the latter is incorporated by standardized load-deflection characteristics of mattress and bed base. Two correction factors were applied to account for differences in shape and contact area of distinct body segments. The resulting shapes were validated by means of spine shape reconstruction based on back surface measurements, showing an average root mean square deviation of 11.9 mm. Furthermore, the distinct mattress–bed base combinations were ranked according to their ergonomic bed score for lateral sleep positions (ESB_L). Rankings based on the predicted spine shapes showed a good correspondence (85 %) with rankings based on the validated spine shapes. Moreover, the two persons for which the superior combination was not correctly identified showed very little difference in ESB_L between the top two combinations.

Although the developed methodology provides an important first step towards objective decision criteria for sleep system selection, some limitations remain present. First, only lateral sleep positions were considered. In fact, healthy sleep is characterized by the presence of multiple position shifts throughout the night [2]. Therefore, it makes sense to incorporate

Table 1
Summary statistics of measured anthropometric data for the entire subject population (n = 15)

<table>
<thead>
<tr>
<th></th>
<th>Mean ± std</th>
</tr>
</thead>
<tbody>
<tr>
<td>Body length [mm]</td>
<td>1749 ± 67</td>
</tr>
<tr>
<td>Body weight [kg]</td>
<td>71.0 ± 8.6</td>
</tr>
<tr>
<td>Shoulder height [mm]</td>
<td>1410 ± 57</td>
</tr>
<tr>
<td>Breast height [mm]</td>
<td>1307 ± 49</td>
</tr>
<tr>
<td>Waist height [mm]</td>
<td>1112 ± 44</td>
</tr>
<tr>
<td>Pelvis height [mm]</td>
<td>847 ± 42</td>
</tr>
<tr>
<td>Shoulder breadth [mm]</td>
<td>440 ± 29</td>
</tr>
<tr>
<td>Breast breadth [mm]</td>
<td>376 ± 31</td>
</tr>
<tr>
<td>Waist breadth [mm]</td>
<td>342 ± 14</td>
</tr>
<tr>
<td>Pelvis breadth [mm]</td>
<td>392 ± 21</td>
</tr>
<tr>
<td>Shoulder width [kg]</td>
<td>12.1 ± 1.6</td>
</tr>
<tr>
<td>Breast width [kg]</td>
<td>10.3 ± 1.3</td>
</tr>
<tr>
<td>Waist width [kg]</td>
<td>9.4 ± 1.1</td>
</tr>
<tr>
<td>Pelvis width [kg]</td>
<td>18.3 ± 2.2</td>
</tr>
</tbody>
</table>

Table 2
Descriptive statistics of the predicted and validated ergonomic bed scores.

<table>
<thead>
<tr>
<th></th>
<th>Simulation</th>
<th>Validation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean</td>
<td>5.20</td>
<td>6.09</td>
</tr>
<tr>
<td>St. dev.</td>
<td>1.09</td>
<td>1.83</td>
</tr>
<tr>
<td>Max.</td>
<td>8.07</td>
<td>9.99</td>
</tr>
<tr>
<td>Min.</td>
<td>3.70</td>
<td>3.37</td>
</tr>
</tbody>
</table>

Figure 6: Boxplots of the root mean square deviations (RMSD) between predicted and measured spine shapes for the nine possible mattress–bed base combinations.
spine shape assessment in supine and prone positions as well. A second limitation involves that body support is not the only factor that determines sleep comfort. Several high quality studies have shown that the thermal environment might affect sleep initiation and sleep depth [19, 20]. Since the bed microclimate is highly affected by the thermal properties of the sleep system, information on thermal insulation and vapor permeability should be considered as well when choosing a sleep system. Therefore, future work will look at characterizing the thermal interactions between the human body and the sleep system that take place during sleep [23].

In conclusion, the presented work provides a first step towards the development of objective criteria to evaluate support properties of sleep systems on a personalized level. It was shown that spine shape could be predicted by combining a personalized anthropometric model with load-deflection behavior of mattress and bed base.

References


