



Passenger Car' Sitting Posture Prediction Research

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Abstract. Sitting posture as the input of the design of Automobile seat face, makes a huge effect on the overall seat comfort. The current software for sitting posture calculation is based on the SAE manikin and European driving habits for sitting, which is much different from the Chinese one. This paper firstly analyzes the key factors that affect the driving posture, and defines the input of the prediction model, and then predict the position of eye points and H point in different vehicles for different drivers, with the step-wise regression method. A inverse kinematic method is applied to solve the posture of torso, upper limbs and lower limbs to obtain the overall sitting posture. Finally, the accuracy of the model is verified by a comparison between the measured data and the predicted one.

Keywords: Sitting posture · Influencing factors · Stepwise regression · Inverse kinematics · Model optimization

1 Introduction

Driving posture definition is the key for automotive seat design. After the interior layout is finished, the digital human model is used to define the driving angles, based on which, the supporting forces of the seat cushion and back are designed, and the seat foam then can be designed. Therefore, the accurate posture definition is crucial to the overall seat comfort.

The sitting posture is not only influenced by the anthropometry, but also is directly related with the driving habits in different countries [1]. A significant difference between the Chinese anthropometry and the European and American ones exists, which leads to a big discrepancy also, therefore, it is necessary to establish the driving posture prediction model based on the Chinese anthropometry.

This paper builds the sitting posture prediction model based the Chinese anthropometry, using stepwise regression and inverse kinematics methods [2–4], and realizes the purpose that predict sitting postures of different drivers in different seats, which also could provide automobile engineers the posture defining and checking tools.

2 Experimental Scheme

The stratified sampling is used to recruit 1800 drivers from 6 cities of China, in which female subjects are 1/3 of the male ones, and their ages are in uniform distribution among 20–60.

There are 3 passenger cars for measuring, which are coded as A, B and C. Every subject’s sitting postures in 3 cars are measured respectively, in order to research on the posture difference among drivers in different cars, as Fig. 1(a) shows [5].

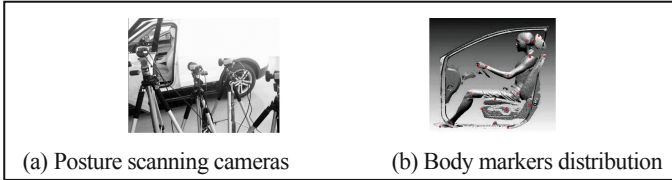


Fig. 1. Posture scanning facility and body markers

The subjects enter the car’s driving position after their bodies are marked with reflective makers, just as Fig. 2(b) shows. They are guided to adjust seat to the most comfortable status in their mind, after that, subjects’ sitting postures are measured and recorded, as Table 1 shows.

Table 1. Sitting posture angles list

A1	A2	A3	A4	A5	A6	A7	A8	A9
Torso angle	Hip angle	Knee angle	Ankle angle	Upper arm angle	Elbow angle	Wrist angle	Heel angle	Thigh angle

3 Posture Influencing Factors

This paper explores impact from gender and vehicle to sitting postures using repeated measures ANOVA. The Bonferroni correction is adopted to make multiple comparisons among different factors to investigate the simple main effect if interaction effect exists between the two factors. Effects are considered as “significant” when $p < 0.05$.

The stepwise regression model is used to explore the relationships between the postural data and the age and anthropometric measurements. Define the entry and removal of using probability of F as 0.05 and 0.1 respectively. The analysis finds that gender and vehicle both have a significant effect on the posture [5].

The stepwise regression analysis between the posture angles and anthropometry and gender needs to standardize all three types data, and results are demonstrated as Table 2.

Table 2. Stepwise regression results

Vehicle	Factors	Postural angles								
		A1	A2	A3	A4	A5	A6	A7	A8	A9
A	AGE	0	-0.082	0.072	0	0	0	0	0	0.043
	Stature	0	-0.535	0	-0.133	0	0	0	0	0.614
	WEI	-0.205	-0.116	0	0.105	0.134	0	-0.085	0.023	0
	SH	0.065	0.133	0.063	0	0.127	0.26	-0.117	0	-0.213
	TRH	0	0	-0.252	0	0	-0.06	0.104	0	0
	SAH	0	0.086	0	0	0	0.102	-0.132	0	0
	ATO	0	0	0	0	0	0	0	0	0
B	AGE	0	0	0	0.055	0	0	-0.054	0	0
	Stature	0.287	-0.259	0	0	0	0.211	0	0	0.543
	WEI	-0.296	-0.142	0.097	0	0.147	0	-0.107	0	-0.07
	SH	0	0	0	0	0	0	0	0	0
	TRH	-0.139	-0.117	-0.256	0	0	-0.149	0	0	0
	SAH	0	0.089	0.066	0	0.088	0.242	-0.13	0.105	-0.112
	ATO	0	0	0	0	0	-0.08	0	0	0.075
C	AGE	0	0	0	0	0	0	0	0.105	0
	Stature	-0.019	-0.401	0	-0.221	0	0	0	0	0.539
	WEI	-0.284	-0.134	0.075	0	0.167	0	-0.071	-0.148	-0.07
	SH	0	0	0	0	0	0	0	0	0
	TRH	0	0	-0.114	0	0	0	0	0	0
	SAH	0	0.109	0	0.114	0.071	0.142	-0.113	0	-0.133
	ATO	0	0	-0.097	0	0	0	0	0	0

The anthropometric measurements include: stature, weight (WEI), sitting height (SST), trochanter height (TRH), sitting acromion height (SAH), acromion to olecranon length (ATO).

Table 2 tells that Stature, WEI, SH, SAH and also vehicle are the main factors influencing postural angles, which are selected as the inputs to the prediction model.

4 Posture Prediction Model

Driver's anthropometry, seat parameters and vehicle layout make effects on the postures, which are taken as input to the model, as Fig. 2 shows.

Simplification is done to the prediction model to reduce the calculation complexity. The three-dimensional human model is simplified as the two-dimensional one, considering that sitting posture mainly indicates the joint angles. Also, it is deemed that normal driving posture is sagittally symmetric, and hands are assumed to be continuous with the forearms, and foot posture is neglected. The coordinate is simplified as x-z plane without y axis, as only the two-dimensional driving posture is concerned.

The eye and H point location regression equations are as Table 3 shows.

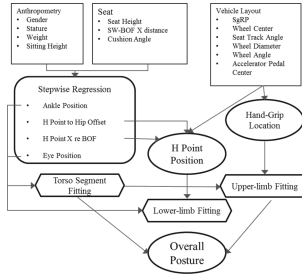


Fig. 2. Posture prediction model schematic diagram [1]

Table 3. Posture regression equation

Variables	Stature	Sitting height	H30	SW-BOFX	Cushion angle	Adjusted R ²	RMSE
Hip X re BOF	0.4701	-429.8	-0.173	0.4511	-1.04	0.67	36.3
Hip-to-eye angle	0.00638		115.73	0.0188	0.11	0.21	4.4
Eye X re BOF	0.5758	913.4	-0.1514	0.5847		0.71	51.1
Eye Z re AHP	0.3134	615.7	1.04	0.0288		0.79	22.3
Eye X re hip	0.1177	1350.2		0.155	1.22	0.31	37.9
Eye Z re hip	0.3324	665.4		-0.0221		0.57	25.7
Ankle X re BOF	0.06	458.1	0.175	0.147	1.5	0.28	20.3
Ankle Y re A pedal	-0.047					0.11	25.4
Ankle Z re AHP	0.0341		0.1312		0.47	0.21	18.4
Knee angle	-0.0066	57.3	-0.0341	0.0785	-0.61	0.34	10.2
Head angle	0.00922	141.5				0.04	12.3
Neck angle	-0.01214		0.0114			0.05	11.2
Thorax angle	0.01214	46.1		0.00998		0.04	8.2
Abdomen angle	0.011	190.5		0.0314		0.11	8.5
Pelvis angle	0.0112	101.2		0.0181	0.41	0.05	11.4

The upper extremity and lower limb fitting are calculated using inverse kinematic method after the hip and eye locations are obtained. This paper solve the inverse kinematic with MATLAB Robotics toolbox.

The postural data of a female subject and a male subject are collected respectively to verify the model accuracy. The measured and predicted results are shown in Table 4.

Table 4. Comparison of measured and predicted results

Posture	Subject					
	Male			Female		
	Measured	Predicted	Error	Measured	Predicted	Error
Torso angle	23.755	22.566	5.01%	18.8326	19.411	3.07%
Hip angle	101.254	106.78	5.46%	95.6201	104.36	9.14%
Knee angle	113.33	112.554	0.68%	109.09	110.22	1.04%
Forearm angle	19.4491	22.563	16.01%	56.4071	61.47	8.98%
Elbow angle	108.362	112.45	3.77%	143.651	147.225	2.49%
Thigh angle	12.2552	13.41	9.42%	11.0793	12.177	9.91%

The comparison results show that all the posture prediction error are all below 10% except for the male forearm posture.

5 Conclusion

It is found that driver’s characteristics, vehicle design make a significant effect on the sitting posture. Take the identified effect factors as inputs to the model to predict eye, hip location, upper and lower limb fitting.

A comparison concludes that the model could predict Chinese driver’s sitting postures well except the male forearm’s error exceeds 10%.

The study was approved by the Logistics Department for Civilian Ethics Committee of China.

All subjects who participated in the experiment were provided with and signed an informed consent form.

All relevant ethical safeguards have been met with regard to subject protection.

Compliance with Ethical Standards. The study was approved by the Logistics Department for Civilian Ethics Committee of China Auto Information Technology Co., Ltd.

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