



Düzce Üniversitesi Bilim ve Teknoloji Dergisi

Araştırma Makalesi

Development of a 3D Virtual Welding Simulator Using Weld Bead Created by Voxelization Technique

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ABSTRACT

In this study, we developed and implemented a cost-reducing, real-time virtual welding simulator to train welder candidates. In order to make a real-time welding simulation, a three-dimensional weld bead form was designed. We used a parabola as the basic bead slice shape, considering the similarity between the parabola and the bead slice. During the welding process, the parameters of the weld bead shape are calculated at each time step using an artificial neural network. This network determines the shape of the weld bead and the depth of penetration, based on inputs received from the sensor device that tracks the motions of the torch. After the parabola's parameters have been determined, the voxel map and corresponding hash-based octree data structure are generated in real-time. By using the voxelized data, a weld bead isosurface consisting of triangles is reconstructed with a marching cubes algorithm allowing us to generate more realistic weld seam shapes. We used multi-threaded programming for voxelization and isosurface extraction to reduce the computation cost on high-resolution virtual scenes. The isosurface extraction times for different thread counts and also a feature comparison with other simulators in the literature are shown in this paper.

Keywords: welding simulator, virtual reality, weld bead, weld seam, voxel.

Vokselleştirme Tekniği ile Oluşturulan Kaynak Dolgusunu Kullanan 3B Bir Sanal Kaynak Simülatörü Geliştirilmesi

ÖZ

Bu çalışmada, kaynakçı adaylarının eğitiminde kullanılmak üzere gerçek zamanlı ve maliyeti düşük bir sanal kaynak simülatörü tasarlanıp geliştirilmiştir. Gerçek zamanlı bir kaynak simülasyonu yapmak için öncelikle üç boyutlu bir kaynak dikiş formu tasarlanmıştır. Parabol ve kaynak dikiş arasındaki benzerlik göz önünde bulundurularak temel kaynak dikiş formu olarak parabol kullanılmıştır. Kaynak işlemi sırasında, yapay sinir ağı kullanılarak her zaman adımında kaynak dikiş şeklinin parametreleri hesaplanır. Bu ağ, torcun hareketini izleyen sensör cihazından alınan girdilere dayalı olarak kaynak dikişinin şeklini ve derinliğini belirler. Parabolün parametreleri belirlendikten sonra, gerçek zamanlı olarak voksel haritası ve karşılık gelen sekizli-ağaç veri yapısı oluşturulur. Vokselleştirilmiş veriler kullanılarak üçgenlerden oluşan kaynak dikiş eş-yüzeyi, daha gerçekçi kaynak dikiş şekilleri oluşturmamızı sağlayan yürüyen küp algoritması ile yeniden yapılandırılmıştır. Yüksek

çözünürlüklü sanal sahnelerde hesaplama ve işlem maliyetini düşürmek amacıyla vokselizasyon ve eş-yüzey çıkarma işlemleri için çok iş parçacıklı programlama tekniği kullanılmıştır. Bu çalışmada, farklı iş parçacıkları için eş-yüzey çıkarma süreleri gösterilmiş olup geliştirilen simülâtörün literatürdeki diğer simülâtörlerle karşılaştırması da sunulmuştur.

Anahtar Kelimeler: kaynak simülâtörü, sanal gerçeklik, kaynak dolgusu, kaynak dikişi, voksel.

I. INTRODUCTION

Welding is an economical joining method used worldwide in many applications, such as shipyards, the automotive industry, steel construction structures, bridges, and machine production. Considering technological concepts such as cost and safety for optimum production, it is required that welders have suitable training to obtain their qualifications and efficiencies. Welder education is divided into technological knowledge and hands-on skill training. Although technological knowledge and experience are essential in welding, hands-on skill training is the most crucial parameter in welding joints [1-2]. This is because weld joints are performed with hand motions and angles. Other important welding factors are weld materials, wire size, shielding gas flow rate, wire feed rate, welding speed, and welding voltage [3-5].

During training, a welder candidate can waste many test parts while working on developing his hand skills. This process is repeated often during a welder's training, and test parts cannot be used again. Because of this, the total cost of used test parts becomes high. If skills and experience are gained via welding simulators before actual welding, this can lead to huge savings of energy, time, materials, security and cost [6-7].

Research in welding simulators started to appear in literature at the beginning of the 1990s. [8] investigated how 2-dimensional (2D) welding simulators affected welder education. They discovered that studying with welder simulators decreased education time and cost. They advised using virtual training before the welder workshop education stage. Top [9] analyzed welding simulations in training. He observed that the students who were educated on welding simulators using MIG (Metal Inert Gas) and TIG (Tungsten Inert Gas) welding methods had better performance during real welding operations, with better welding quality and reduced scrap parts.

Oz et al. [10] divided welding simulators into two categories: older and newer generation. One of the new generation welding simulators, which uses virtual reality (VR) hardware and software technology, is set up for 3D interactive MIG welding [11]. The user can adjust the parameters with a graphical interface, then perform the welding operation and see the results. Another MIG welding simulation system has the combination of a haptic device, a head-mounted display (HMD), a 6-degree-of-freedom tracing system, and a speaker. A torch is connected to the haptic device to track the torch position. It is a real-time simulator and uses an artificial neural network to obtain a welding seam form [12]. White et al. [13] added heat distribution to Fast's work to calculate the penetration. Some simulators such as the VRTEX 360 [14], Soldamatic AR welding simulator [15], and an Electric Arc Simulator [16] use optical trackers for position detection.

This study aims to develop a virtual welding simulator with an ideal design and a working plan with an optimum cost. This welding simulator can be used for MIG and TIG welding applications, both widely used welding methods in various industries. In order to obtain better results in virtual welding applications, all required parameters in welding are determined and stored in a database. A software system is developed to supply interaction with virtual welding models. An HTC Vive controller is used as an input unit, and it perceives the position and inclination of the hand motions of the user. A Vive headset is a head-mounted display used as an output unit. This virtual welding simulator allows welder applicants to perform many applications with different welding parameters. Thus, many visual effects such as feed rate [17], sound, light, and welding seam form met in the real welding applications and are

simulated in a shape close to the real welding. After welding, errors made by the candidate can be understood with the performance evaluation system and improved for better performance.

The remainder of this paper is organized as follows: The developed virtual welding simulator is presented in Section 2. Then, we described how we created the weld seam in detail in Section 3. Experimental results and comparisons of our virtual welding simulator system are given in Section 4. Last, we conclude our work in Section 5.

II. THE DEVELOPED SIMULATOR

For the positive contribution of virtual reality to training [18], a virtual welding simulator was developed in this study to train welder candidates. Figure 1 shows the overall structure of our virtual welding simulator system. Firstly, the welder selects the initial welding parameter values such as current and voltage. Then, the user starts the welding process, and the simulator receives data from input devices. According to the data received from these devices, the virtual welding environment is constantly updated to increase the interaction. The reconstructed virtual scene is sent to the output devices, such as the HMD and 3D monitor.

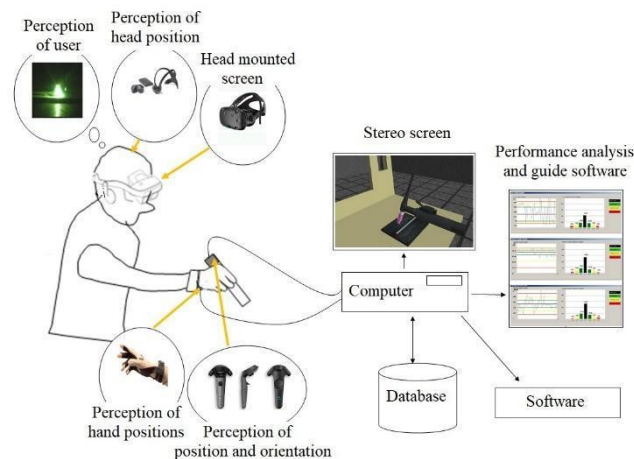


Figure 1. The general structure of our virtual welding simulation system

The system consists of five main modules: a graphics module, an input/output (I/O) module, a weld bead parameter decision module, a performance analysis module, and a sound module. Figure 2 shows the architecture of our virtual welding simulator. A virtual welding simulation software was developed with the utilization of the C++ object-oriented programming language, OpenGL graphics library, and Open Audio Library (OpenAL). The developed simulator uses special virtual reality devices: a data glove, an HTC Vive controller, and an HTC Vive headset. The system receives the x, y, z location and roll, pitch, yaw orientation values of the user's hand by the HTC Vive controller. In the weld bead parameter decision module, the motion data is evaluated by an artificial neural network and the parameters of the weld bead are obtained. After processing the data, the system creates and shows the virtual welding scene on the HTC Vive headset. At the same time, the sound engine renders 3D positional audio. After the welding application, the weld seam is analyzed by the performance analysis module.

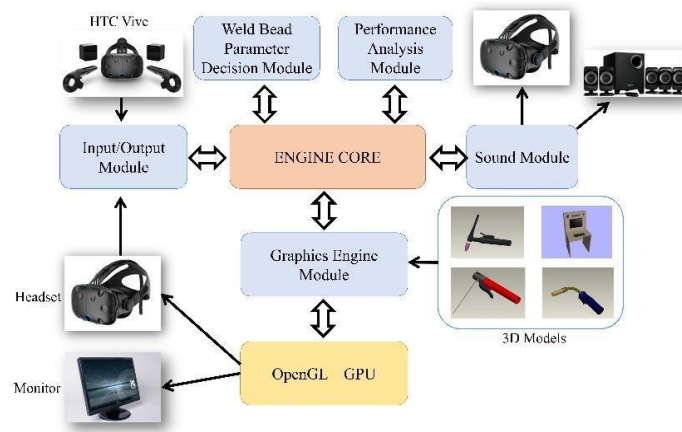


Figure 2. Architecture of our virtual welding simulator

A. INPUT/OUTPUT MODULE

Computer I/O devices are designed to work while sitting on a chair or in a particular sitting position. However, a virtual reality system needs specially designed I/O devices, such as a data glove, a Flock of Birds, and an HMD, to track human motion [19-20]. Our simulator uses an HTC Vive controller device to track the welder's hand motions.

The Cyberglove detects hand gestures with its 18 sensors. These sensors' frequency is 150 Hz, which is high enough to catch every motion of the user's hand.

The welder can see the virtual world through an HTC Vive headset. The virtual welding environment is updated constantly according to the user's head motions, which the Vive headset tracks. In this way, when a user turns his head, the virtual camera moves accordingly. A welding platform was built to increase the interaction between the welder and the virtual environment. Also, an auto stereo 3D screen was mounted on this platform to show viewers the welding process.

The HTC Vive headset has two OLED panels that show the virtual scene through two closely placed cameras. Each of these panels has a display resolution of 1080x1200, which is quiet enough for our simulator. These two displays are placed right before each eye so that users can see the virtual world as they see the real world. When the user wearing a Vive headset moves his/her head, sensors on the Vive headset detect this motion and feed this information to simulator software. The HTC Vive headset uses dozens of infrared sensors to determine the headset's position in a space. Also, the headset has a gyroscope, a proximity sensor and a G-sensor. After receiving head motion information, the virtual software updates the virtual camera's position and orientation, making the user part of the virtual world.

In the simulator, an HTC Vive controller is used to represent a torch. In the simulation, HTC Vive controllers track the user's hand motions with their sensors. When the user holding the controller moves his hand, the torch in the virtual environment moves simultaneously. This way, the user gets the feeling of controlling an actual torch. An HTC Vive controller has 24 infrared sensors, and also to aid its motion tracking ability, it has an InvenSense's MPU-6500 6 DOF sensor. MPU-6500 has a 3-axis accelerometer and a 3-axis gyroscope.

B. GRAPHICS MODULE

Our simulator uses a graphics engine to draw and update the virtual scene. The graphics engine in the simulator is written in C++ programming language using the OpenGL library. OpenGL is a set of low-level Application Programming Interfaces (APIs) that provide access to a graphics card's capabilities, enabling programs to provide realistic 3D graphics. We use the OpenGL library to send instructions to the graphics card and provide high-performance hardware-accelerated multimedia support.

We designed 3D welding models that are based on real objects used in real welding operations and placed them into the virtual world of the simulator. 3D welding models are composed of polygons. If a model detail is high (close to the real object), then its polygon counts are high and may require more processing power to draw this model [21]. The first step was designing 3D virtual models of real objects using Pro/Engineer software. These objects are torches, metal parts, and a table. After this step, the necessary texture is applied to the models. Some of the 3D objects, such as weld beads, are created dynamically in real-time by the VR graphics engine.

C. PERFORMANCE ANALYSIS MODULE

Our expert system is used to evaluate and analyze welder performance. Expert systems are one of the common artificial intelligence methods for solving engineering problems like a human expert. Our expert system model rules [22] were created by four experts. These experts were a welder trainer, two mechanical engineers, and a metallurgist. By using this module, virtual welding experiments performed by welder candidates can be evaluated, and the results are shown by scoring or graphics. Based on the rules determined by the welding experts, the trainee's performance score is graded by a comparison of the recorded values of each welding parameter to the ideal values. Any welding operation performed previously can be reconstructed at any time using the recorded data in the database. In this way, welder candidates' personal development can be monitored for current or retrospective welding applications.

D. WELD BEAD PARAMETER DECISION MODULE

Hand motion data received from the I/O module is evaluated by the artificial neural network, and the parameters of the weld bead are obtained in the weld bead parameter decision module. The details of this module are explained in Section 3, titled "Forming the Weld Seam".

E. SOUND MODULE

The sound module in the simulator is written in C++ programming language by using the OpenAL library. OpenAL sound API is used to create sound effects because it is designed for efficient rendering of 3D positional audio. The API's purpose is to allow a programmer to position audio sources in a 3D space around the user. Sound effects recorded from real welding operations are positioned in the virtual environment so that a user in the virtual environment feels as though they are in a three-dimensional world.

III. FORMING WELD SEAM

The weld seam is a series of deposits of filler metal in the welding process. When performing the virtual welding application, the sensor on the torch detects the user's hand movements. The parameters of the weld bead are calculated using this hand motion data received from the sensor, and the voxel map of the weld shape is generated in real-time. Then, the hash-based octree data structure approach [23] is applied to the dataset to manipulate the corresponding voxels easily. The hash-based octree structure reduces the memory space, computation cost, and tree traversal time. After obtaining the voxel map, the corresponding isosurface is reconstructed during the welding process.

The determination of the weld bead basic shape, the prediction of output parameters, and the voxelization process are described in detail in the following sections below.

A. DETERMINATION OF THE WELD BEAD BASIC SHAPE

In the literature, welding bead studies [24-39] show that the basic welding bead shape appears in Figure 3. Therefore, we decided to use this general bead shape in this study.

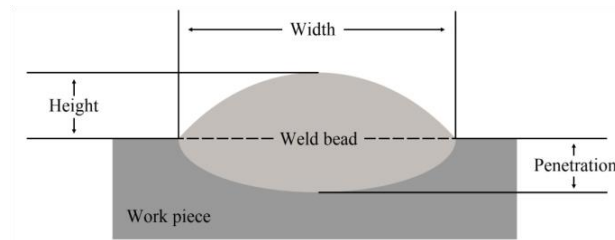


Figure 3. Weld bead shape and parameters

In order to obtain the height, width, and penetration values to be used to obtain the model shown in Figure 3, the parameters used during the welding process were examined. Tab. 1 lists the input and output parameters used in academic studies involving MIG and TIG sources in the literature [40-57]. When the studies in Table 1 are analyzed and the suggestions from our expert system mentioned in section 2-C are taken into consideration, it is seen that welding speed, current, arc length, and travel angle directly affect the welding geometry for MIG welding; and welding speed, current, and arc-distance directly affect the welding geometry for TIG welding.

Table 1. Input/Output parameters for MIG and TIG welding

Welding Type	Reference	Input Parameters						Output Parameters		
		Welding Speed	Current	Voltage	Arc-Length	Travel Angle	Wire Speed	Penetration	Width	Height
MIG	[40]	√	√	√				√	√	√
MIG	[41]	√		√				√	√	√
MIG	[42]	√	√		√	√		√	√	√
MIG	[43]	√	√	√				√	√	√
MIG	[44]	√	√	√					√	
MIG	[45]	√	√	√				√		
MIG	[46]	√	√	√				√		
MIG	[47]	√	√	√				√		
MIG	[48]	√	√					√	√	√
TIG	[49]	√	√					√	√	√
TIG	[50]	√	√	√			√	√	√	√
TIG	[51]	√	√		√		√	√	√	√
TIG	[52]	√	√		√			√		
TIG	[53]	√	√		√			√	√	√
TIG	[54]	√	√		√		√		√	√
TIG	[55]	√	√	√					√	√
TIG	[56]	√	√					√	√	√
TIG	[57]	√	√					√	√	

When Figure 3 and Table 1 are evaluated together, and when we examined how to create Fig. 3 with the parameters obtained from Table 1, it was thought that this shape could be created with 2 parabolas. Also, Chambers and Wu [58-59] stated that the welding seam section resembled a parabola. They use a welding seam volume by rotating the y-axis with the centre of the parabola (0,0). Mavrikios [11] also used an ellipsoid in his study to find welding seam volume.

The parabola is known in algebra as the graph of the second-order functions, as shown in Eq. 1.

$$y = a(x - r)^2 + k \tag{1}$$

Here, a denotes the direction of the parabola, while the point (r, k) indicates the peak of the parabola. The graph of this equation is shown in Figure 4-a. In this study, the parabola form in Figure 4-b is used because of the resemblance of the weld seam section to a parabola. Here, h is the height of the weld seam section, and w is the width of the weld seam section. When the welding application is performed with the virtual welding simulator, the parabola equation is calculated using the input values w, h and p . An actual weld bead slice shape generated by our simulator is shown in Figure 4-c.

If parabola values in Figure 4-b are substituted in Eq. 1, then $r = 0$ and $k = h$, and Eq. 2 is:

$$y = a(x)^2 + h \tag{2}$$

Eq. 3 is created when we want to reach the value a :

$$x = \frac{y-h}{a} \tag{3}$$

Again, if $x=w/2$ value of parabola in Figure 4-b is written in Eq. 3 for $y = 0$, then value of a is obtained in Eq. 5:

$$a = \frac{0-h}{(w/2)^2} \tag{4}$$

$$a = \frac{4h}{(w)^2} \tag{5}$$

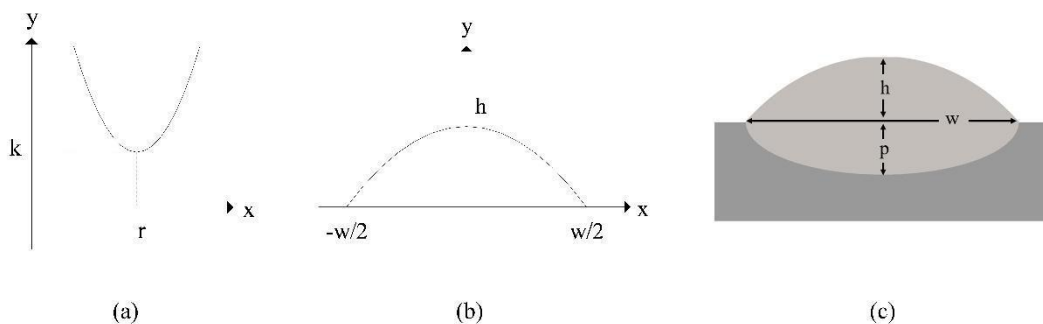


Figure 4. Parabola to weld bead slice simulation. (a) Graph of a parabola (b) Parabolic shape decided for weld bead (c) Weld bead slice shape and its parameters

B. PREDICTION OF OUTPUT PARAMETERS BY ARTIFICIAL NEURAL NETWORKS

TrainLM was used as an educational function when planning the artificial neural network structure because it works effectively in network training [60-62]. In determining the transfer function, it was found that the LogSig() and TanSig() transfer functions produced close results, but the LogSig() function gave the best result.

The schematic representations of the neural networks for the MIG and TIG weldings are shown in Figure 5. The MIG welding data used for network training, verification, and testing were taken from Sreeraj's 32 real MIG welding applications [42]. The TIG welding data used for network training, verification, and testing were taken from Esme's 16 real TIG welding applications [53]. 70% of the data was used for training, 15% for verification, and 15% for testing. For the MIG welding application, the input layer consists of current, welding speed, arc-length, and travel angle variables. The input layer consists of current, welding speed, and arc-length variables for the TIG welding application. The number of hidden layers and the number of process elements (neurons) in each hidden layer were investigated by a trial method and it was determined to be 14 neurons for MIG and six neurons for TIG welding in one layer.

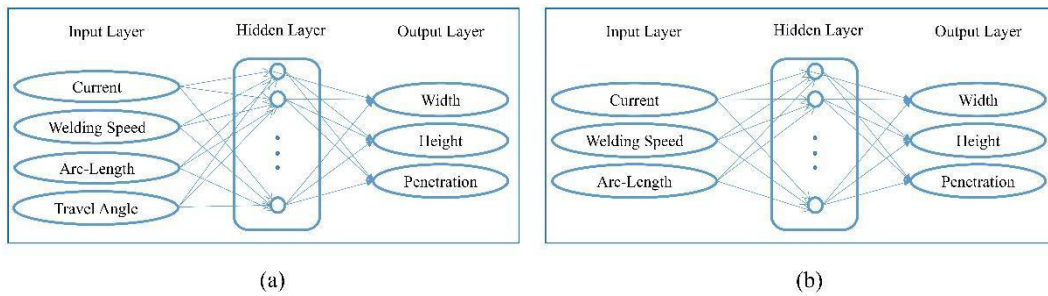


Figure 5. Schematic representation of the artificial neural network used for the prediction of weld seam dimensions (a) MIG welding (b) TIG welding

C. VOXELIZATION

One of the major features of voxel data is that it can contain physical properties such as heat, temperature, colour, density, and hardness. Therefore, we preferred to use a volumetric voxelization approach to represent the weld seam data. We used a triangular mesh model for a 3D weld bead image, and to obtain this model, a voxel map [63] of the weld bead was needed in the background. In order to examine the weld seam, surface, and volume, voxel information was needed. First, we voxelized our bead's parabola model and then we used a flood-filling algorithm [64] to fill the model's interior volume. We used optimized hash tables to store the volumetric voxelized weld seam data.

Voxel shapes are not suitable for representing 3D models because of aliasing problems. For this reason, we used a marching cubes algorithm [65-67] on voxel-based octree data to reconstruct the weld seam surface.

The processing time of interior volume filling was reduced by using the multi-threaded programming approach [21, 68]. Tab. 2 gives the voxelization times of the weld seam model. The voxelization times in this table are calculated according to the parabolas, resolutions, and thread numbers. The results given in this table were obtained with a PC with an Intel I7 5700HQ 2.7 GHz processor with four physical cores and eight logical threads. When the table is examined, it is seen that multi-threaded programming performs better on high-resolution models.

After filling the interior volume, the model's surface was reconstructed using the Marching Cubes algorithm, the most popular isosurface extraction algorithm. The Marching Cubes algorithm method creates triangles by generating a lookup table from volumetric cube data. It is the most preferred technique because of its ease of application [66, 69-73].

Table 2. Voxelization times for different thread counts for a sample parabola

Resolution	Number of Voxels	Process Time of Parabola Voxelization(ms)			
		Number of Threads			
		1	2	4	8
256x256x256	140	0,232	0,171	0,154	0,142
512x512x512	280	0,505	0,328	0,296	0,268

III. EXPERIMENTAL RESULTS AND COMPARISON

In this study, we developed a virtual welding simulator. The virtual welding simulator and a sample of welding are shown in Figure 6.

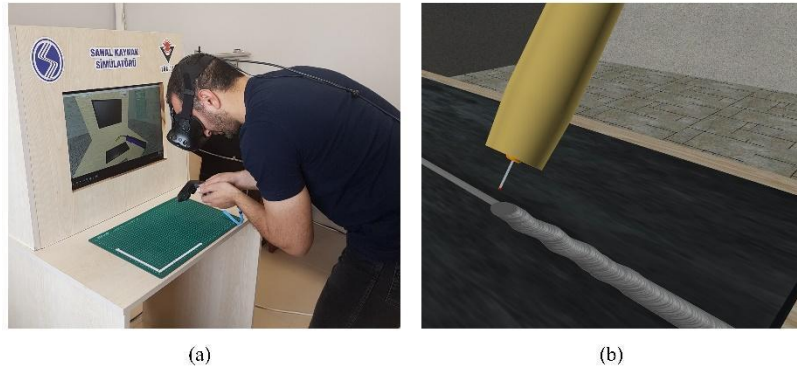


Figure 6. *Welding with virtual welding simulator (a) Developed virtual welding simulator (b) Welding sample with virtual welding simulator*

Figure 7 (a) and (b) show a parabolic and voxelized view for a weld seam sample. This illustrates how the welding seam gains volume. During welding operations, a weld seam is created in real-time, and to enhance the sense of reality, we created welding sound, light, and sparks. After the virtual welding application, voxelized welding seam forms can be evaluated. Figure 7-c and 7-d show a voxelized seam form after welding operation.

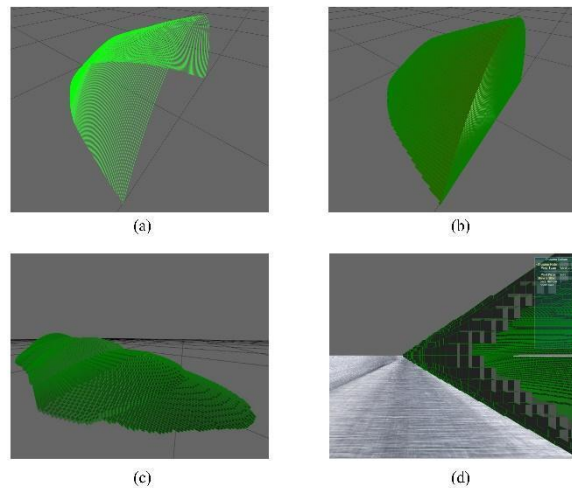


Figure 7. *Parabolic and voxelized weld seam views (a) parabolic view, (b) voxelized view, (c) top view, (d) side view*

Multi-threaded programming makes welding seam drawing even more suitable for real-time applications. Considering that real-time images should have a 30-60 frame per second display, one frame should be drawn under periods of 16.7 ms - 33.3 ms. As shown in Tab. 3, the total drawing time for one parabola meets the real-time image criteria.

Table 3. Total Process Time of a Parabola

Resolution	Number of Voxels	Total Process Time of Parabola (ms)			
		Number of Threads			
		1	2	4	8
256x256x256	140	3,00	1,84	1,22	1,12
512x512x512	280	5,75	3,54	2,38	2,10
1024x1024x1024	560	11,78	7,10	4,76	4,27

Our welding simulator has a 3D view concept that makes the user feel as though they are genuinely welding. The welding operation has boundaries such as welding area and arc-length. Trainees use stereo HMD to immerse themselves in the welding environment. The audience and welder trainer see the welding operation on an auto stereo screen in 3D.

Most of the analyzed simulators, such as the VRTEX 360 [14] and Soldamatic AR welding simulator [15] in the literature, use third-party commercial graphics libraries, however, we used only free libraries and created our own graphics engine. Thus, the total cost of the simulator was decreased. Using a haptic device [12] narrows the field of work, and the data-glove [11] tracks finger positions that are not directly necessary for torch position. Using cameras [13] for position detection requires additional calculations compared to the HTC Vive, which needs no additional computation. The absence of particular performance scoring reduces the benefits of a simulator. Also, many simulators in the literature implement only MIG welding. A detailed comparison of the analyzed simulators is given in Tab. 4.

Table 4. Comparison of analyzed simulators

Simulator	Welding type	Graphics support/engine	3D view	Weld puddle	HMD in helmet	Position detection	Welding sound	Performance scoring
Current Study	MIG, TIG,EA	own	√	√	√	HTC Vive	√	√
[11]	MIG	Infinite Reality 2	√	√	√	Data glove	X	√
[12]	MIG	Endea VR	√	√	√	Haptic device	√	X
[13]	MIG	own	√	√	√	Cameras	√	X
[14]	MIG	VRSIM	√	√	√	Optic tracker	√	√
[15]	MIG, TIG	Unknown	√	√	√	Optic sensors	√	√
[16]	EA	Unity 3D	√	X	X	Optic sensors	√	X

When the *Welding type* column of Table 4 is examined, it is seen that the developed simulator performs both MIG and TIG welding, and there is only one reference with this feature, but in this study [15], it is not mentioned how the graphics engine was created.

When the column titled *Graphics engine* in Table 4 is examined, it is seen that there is only reference [13] which creates its own graphics engine like us. Although coding a graphics engine is a long and laborious task, coding our own engine has advantages in terms of mastering the simulator.

The 3D view of the simulation, the ability to show the weld puddle, the use of the HTC Vive technological device, and the integration of welding sound into the simulator reveal the difference of the proposed study from other simulators. In addition, having performance indicators is the most important indicator that this simulator is an educational simulator that can perform "measurement and evaluation".

Our developed virtual welding simulator system has been tested on welder trainees at Sakarya University. Figure 8 shows the average scores of two different groups of 40 users for each welding task. The scores are marked by the course instructors. Regardless of ability, an increase in users' performance levels and achievement was observed. Also, we observed how candidates' welding skills changed from using our simulator. The results showed that trainees effectively improved their welding capabilities [22].

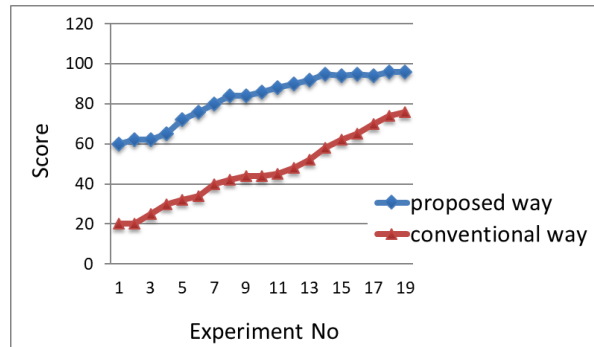


Figure 8. Real welding operation scores of trainees [22]

V. CONCLUSIONS

When a welder candidate uses a welding simulator before a real welding operation, waste in the form of energy, time, and materials is reduced. Also, a welder can work in an environment with no risk of electric shock, burns, or eye burns. The welders are also prevented from having psychological symptoms such as shyness, despondency, and loss of self-confidence.

In this study, a virtual welding simulator was developed to train welder candidates. The work is based on welding simulation without a need for a real welding environment. Trainees can learn welding techniques or practice their skills in safe conditions without the risk of injury and can complete more practice welds in a short period for both MIG and TIG welding types.

In this study, the hand motion data received from I/O devices were examined by the weld bead parameter decision module. We calculated the necessary parabolas that represented the weld bead, and a voxel map was created from these parabolas. A hash-based octree data structure was created to control the corresponding voxels. Then, the triangular mesh model representing the weld seam was constructed employing the hash-based octree data structure using the marching cubes algorithm. As the welding process continues, the hash-based octree structure is updated, and the triangular mesh model of the weld seam is reconstructed in real-time. We used multi-threaded programming for voxelization and isosurface extraction to reduce the processing time on high-resolution virtual scenes. Thus, we generated more realistic weld seam shapes by using voxelized parabolas.

We did not use any third-party commercial libraries, and we created our own graphics engine, decreasing the total cost of the simulator. We made the simulator more economical and commercially available by reducing the cost.

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V. REFERENCES

- [1] G. Karsai, K. Andersen, G.E. Cook, and R.J. Barnett, "Neural network methods for the modeling and control of welding processes," *Journal of Intelligent Manufacturing*, vol. 3, pp. 229-235, 1992.
- [2] S. Teeravarunyou, and B. Poopath, "Computer based welding training system," *International Journal of Industrial Engineering: Theory Applications and Practice*, vol. 6, no. 2, pp. 116-125, 2009.
- [3] K.M. Kanti, and P.S. Rao, "Prediction of bead geometry in pulsed gma welding using back propagation neural network," *Journal of Materials Processing Technology*, vol. 200(1-3), pp. 300-305, 2008.
- [4] Q. Xue, S. Ma, Y. Liang, J. Wang, Y. Wang, F. He, and M. Liu, "Weld bead geometry prediction of additive manufacturing based on neural network," *11th International Symposium on Computational Intelligence and Design*, China, 2018, pp. 47-51.
- [5] P.K. Jayashree, S. Sharma, and N. Shetty, "TIG welding parameters optimization of Al–Si–Mg ternary alloy–SiC powder reinforced composites using Taguchi and RSM techniques," *Cogent Engineering*, vol. 9, no. 1, 2022.
- [6] S.K. Katheria, D. Kumar, T.A. Khan, and M.K. Singh, "Reality based skills development approach in welding technology: An overview," *Materials Today: Proceedings*, vol. 47, no. 19, pp. 7184-7188, 2021.
- [7] C. Papakostas, C. Troussas, A. Krouska, and C. Sgouropoulou, "User acceptance of augmented reality welding simulator in engineering training," *Education and Information Technologies*, vol. 27, no. 1, pp. 791-817, 2022.
- [8] C. Wu, C. Wen, and L. Wu, "A microcomputer-controlled welder training system," *Computers & Education*, vol. 20, no. 3, pp. 271-274, 1993.
- [9] Y. Top, "Using Simulator in Training of Arc Welder," *Sakarya University Journal of Science*, vol. 2, no. 1, pp. 93-98, 1998.
- [10] C. Oz, F. Findik, O. Iyibilgin, U. Soy, Y. Kiyani, S. Serttas et al., K. Ayar, S. Uslu, and Y. Yasar, "Welding simulators: from past to present day," *Metal Dunyasi*, vol. 201, pp. 108-111, 2010.
- [11] D. Mavrikios, V. Karabatsou, D. Fragos, and G. Chryssolouris, "A prototype virtual reality-based demonstrator for immersive and interactive simulation of welding process," *International Journal of Computer Integrated Manufacturing*, vol. 9, pp. 294-300, 2006.
- [12] K. Fast, T. Gifford, and R. Yancey, "Virtual training for welding," *Proceedings of the Third IEEE and ACM International Symposium on Mixed and Augmented Reality*, 2004, pp. 298-299.
- [13] S.A. White, M. Prachyabrued, T.L. Chambers, C.W. Borst, and D. Reiners, "Low-cost simulated mig welding for advancement in technical training," *Virtual Reality*, vol. 15, no. 1, pp. 69-81, 2011.
- [14] <https://www.lincolnelectric.com/en-gb/equipment/training-equipment/vrtex360/pages/vrtex-360.aspx> Accessed 21 May 2018
- [15] *Soldamatic's HyperReal-Sim™: What is it all about?*, December 2022. [Online]. Available: <https://seaberyat.com/en/hyperreal-sim-of-soldamatic-what-is-it-about/>
- [16] V.G. Bharat and P. Rajashekar, "Virtual reality for metal arc welding: a review and design concept," *International Journal of Mechanical Engineering and Technology*, vol. 8, no. 1, pp. 132-138, 2017.

- [17] P.K. Palani, and N. Murugan, "Modeling and simulation of wire feed rate for steady current and pulsed current gas metal arc welding using 317L flux cored wire," *The International Journal of Advanced Manufacturing Technology*, vol. 34, pp. 1111-1119, 2007.
- [18] Ö.K. Kalkan, Ş. Karabulut, and G. Höke, "Effect of virtual reality-based training on complex industrial assembly task performance," *Arabian Journal for Science and Engineering*, vol. 46, no. 12, pp. 12697-12708, 2021.
- [19] C. Oz, and M.C. Leu, "Human-computer interaction system with artificial neural network using motion tracker and data glove," *International Conference on Pattern Recognition and Machine Intelligence*, vol. 3776, pp. 280-286, 2005.
- [20] C. Oz, and M.C. Leu, "American sign language word recognition with a sensory glove using artificial neural networks," *Engineering Applications of Artificial Intelligence*, vol. 24, no. 7, pp.1204-1213, 2011.
- [21] S. Serttas, K. Ayar, G. Cit, and C. Oz, "Multi-threaded application for marching cubes algorithm," *International Symposium On Innovative Technologies In Engineering and Science*, Turkiye, 2014, pp. 821-825.
- [22] C. Oz, S. Serttas, K. Ayar, and F. Findik, "Effect of virtual welding simulator on tig welding training," *Journal of Materials Education*, vol. 37, pp.197-218, 2015
- [23] G. Cit, K. Ayar, and C. Oz, "A real-time virtual sculpting application by using an optimized hash-based octree," *Turkish Journal of Electrical Engineering and Computer Science*, vol. 24, no. 4, pp. 2274-2289, 2016.
- [24] G. Cook, R. Barnett, D. Hartman, and A. Strauss, "Neural network systems techniques in weld modeling and control," *Computer Aided and Integrated Manufacturing Systems Techniques and Applications*, 1997.
- [25] P. Li, M.T.C. Fang, and J. Lucas, "Modelling of submerged arc weld beads using self-adaptive offset neural networks," *Journal of Materials Processing Technology*, vol. 71, no. 2, pp. 288-298, 1997.
- [26] B. Chan, J. Pacey, and M. Bibby, "Modelling gas metal arc weld geometry using artificial neural network technology," *Canadian Metallurgical Quarterly*, vol. 38, no. 1, pp. 43-51, 1999.
- [27] M.I.S. Ismail, Y. Okamoto, and A. Okada, "Neural network modeling for prediction of weld bead geometry in laser microwelding," *Advances in Optical Technologies*, 2013.
- [28] N. Murugan, and V. Gunaraj, "Prediction and control of weld bead geometry and shape relationships in submerged arc welding of pipes," *Journal of Materials Processing Technology*, vol. 168, no. 3, pp. 478-487, 2005.
- [29] D. Jo, Y. Kim, U. Yang, G.A. Lee, and J.S. Choi, "Visualization of virtual weld beads," *Proceedings of the ACM Symposium on Virtual Reality Software and Technology*, 2009, pp. 269-270.
- [30] J. Xiong, G. Zhang, H. Gao, and L. Wu, "Modeling of bead section profile and overlapping beads with experimental validation for robotic gmaw-based rapid manufacturing," *Robotics and Computer-Integrated Manufacturing*, vol. 29, no. 2, pp. 417-423, 2013.
- [31] P.K. Palani, and N. Murugan, "Optimization of weld bead geometry for stainless steel claddings deposited by fcaw," *Journal of Materials Processing Technology*, vol. 190, pp. 291-299, 2007.
- [32] K.N. Gowtham, M. Vasudevan, V. Maduraimuthu, and T. Jayakumar, "Intelligent modeling

combining adaptive neuro fuzzy inference system and genetic algorithm for optimizing welding process parameters," *Metallurgical and Materials Transactions B*, vol. 42, no. 2, pp. 385-392, 2011.

[33] J.E. Pinto-Lopera, J.M.S.T. Motta, and S.C.A. Alfaro, "Real-time measurement of width and height of weld beads in gma processes," *Sensors*, vol. 16, no. 9, pp. 1-14, 2016.

[34] C.M. Horvath, J. Botzheim, T. Thomessen, and P. Korondi, "Bacterial memetic algorithm trained fuzzy system-based model of single weld bead geometry," *IEEE Access*, vol. 8, pp. 164864-164881, 2020.

[35] Y. He, D. Li, Z. Pan, G. Ma, L. Yu, H. Yuan, and J. Le, "Dynamic modeling of weld bead geometry features in thick plate gma based on machine vision and learning," *Sensors*, vol. 20, no. 24:7104, 2020.

[36] S.K. Gupta, S. Mehrotra, A. R. Raja, M. Vashista, and M.Z.K. Yusufzai, "Effect of welding speed on weld bead geometry and percentage dilution in gas metal arc welding of SS409L", *Materials Today: Proceedings*, vol. 18, no. 7, pp. 5032-5039, 2019.

[37] J.H. Park, M. Cheepu, and S.M. Cho, "Analysis and characterization of the weld pool and bead geometry of inconel 625 super-TIG welds," *Metals - Open Access Metallurgy Journal*, vol. 10, no. 3:365, 2020.

[38] A. Siddaiah, B.K. Singh, and P. Mastanaiah, "Prediction and optimization of weld bead geometry for electron beam welding of AISI 304 stainless steel," *International Journal of Advanced Manufacturing Technology*, vol. 89, pp. 27-43, 2017.

[39] A. Sattar, A. Hussain, B. Abbas, M.N. Azam, K. Mehmood, A. Wakeel, and S. Ali, "Optimization of tig welding parameters for Ti-6Al-4V titanium alloy using the taguchi design of experiment," *NUST Journal of Engineering Sciences*, vol. 15, no. 2, pp. 65-77, 2022.

[40] I.S. Kim, K.J. Son, Y.S. Yang, and P.K.D.V. Yaragada, "Sensitivity analysis for process parameters in gma welding processes using a factorial design method," *International Journal of Machine Tools and Manufacture*, vol. 43, no. 8, pp. 763-769, 2003.

[41] M. Shoeb, M. Parvez, and P. Kumari, "Effect of mig welding input process parameters on weld bead geometry on hsla steel," *International Journal of Engineering Science and Technology*, vol. 5, no. 1, pp. 200-212, 2013.

[42] P. Sreeraj, and T. Kannan, "Modelling and prediction of stainless steel clad bead geometry deposited by gma using regression and artificial neural network models," *Advances in Mechanical Engineering*, 2012.

[43] E. Karadeniz, U. Ozsarac, and C. Yildiz, "The effect of process parameters on penetration in gas metal arc welding processes," *Materials and Design*, vol. 28, no. 2, pp. 649-656, 2007.

n

[44] J.S. Son, I.S. Kim, H.H. Kim, I.J. Kim, B.Y. Kang, and H.J. Kim, "A study on the prediction of bead geometry in the robotic welding system," *Journal of Mechanical Science and Technology*, vol. 21, no. 10, pp. 1726-1731, 2007.

[45] S.P. Tewari, A. Gupta, and J. Prakash, "Effect of welding parameters on the weldability of material," *International Journal of Engineering Science and Technology*, vol. 2, no. 4, pp. 512-516, 2010.

[46] H.R. Ghazvinloo, A. Honarbakhsh-Raouf, and N. Shadfar, "Effect of Arc voltage, welding current and welding speed on fatigue life, impact energy and bead penetration of aa6061 joints produced by robotic mig welding," *Indian Journal of Science and Technology*, vol. 3, no. 2, pp. 156-162, 2010.

- [47] B. Das, and B. Debbarma, "Influence of process parameters on depth of penetration of welded joint in mig welding process," *International Journal of Research in Engineering and Technology*, vol. 2, no. 10, pp. 220-224, 2013.
- [48] K. Chen, H. Chen, L. Liu, and S. Chan, "Prediction of weld bead geometry of MAG welding based on XGBoost algorithm," *International Journal of Advanced Manufacturing Technology*, vol. 101, pp. 2283-2295, 2019.
- [49] P.T. Trivedi, and A.P. Bhabhor, "Experimental investigation of process parameters on weld bead geometry for aluminium using gtaw," *International Journal of Science and Research*, vol. 3, no. 5, pp. 803-809, 2014.
- [50] K. Andersen, G.E. Cook, G. Karsai, and K. Ramaswamy, "Artificial neural networks applied to arc welding process modeling and control," *IEEE Transactions on Industrial Applications*, vol. 26, no. 5, pp. 824-30, 1990.
- [51] S.C. Juang, Y.S. Tarn, and H.R. Lii, "A comparison between the back-propagation and counter-propagation networks in the modeling of the tig welding process," *Journal of Materials Processing Technology*, vol. 75, no. 1-3, pp. 54-62, 1998.
- [52] U. Duman, "Modeling of weld penetration in high productivity GTAW," Ph.D. Thesis, Colorado School of Mines, Golden, CO, USA, 2009.
- [53] U. Esme, M. Bayramoglu, Y. Kazancoglu, and S. Ozgun, "Optimization of weld bead geometry in tig welding process using grey relation analysis and taguchi method," *Materials in Tehnologies*, vol. 43, no. 3, pp. 143-149, 2009.
- [54] D.S. Nagesh and G.L. Datta, "Genetic algorithm for optimization of welding variables for height to width ratio and application of ann for prediction of bead geometry for tig welding process," *Applied Soft Computing*, vol. 10, no. 3, pp. 897-907, 2010.
- [55] A. Iqbal, S.M. Khan, and M.H. Sahir, "Ann assisted prediction of weld bead geometry in gas tungsten arc welding of hsla steels," *Proceedings of the World Congress on Engineering*, 2011, pp. 818-821.
- [56] M.S.M. Musthaq and M.M. Batcha, "Predicting the weld bead geometry of gta welding on aisi 202 stainless steel," *International Journal of Engineering & Technology*, vol. 3, no. 2, pp. 2463-2469, 2014.
- [57] A.K. Singh, V. Dey, R.N. Rai, and T. Debnath, "Weld bead geometry dimensions measurement based on pixel intensity by image analysis techniques," *Journal of The Institution of Engineers (India): Series C*, vol. 100, pp. 379-384, 2019.
- [58] T.L. Chambers, A. Aglawe, D. Reiners, S. White, C.W. Borst, M. Prachyabrued, and A. Bajpayee, "Real-time simulation for a virtual reality-based mig welding training system," *Virtual Reality*, vol. 16, no. 1, pp. 45-55, 2012.
- [59] C.S. Wu, M.X. Zhang, K.H. Li, and Y.M. Zhang, "Numerical analysis of double-electrode gas metal arc welding process," *Computational Materials Science*, vol. 39, no. 2, pp. 416-423, 2007.
- [60] M.A. Cavuslu, Y. Becerikli, and C. Karakuzu, "Levenberg-marquardt algoritması ile ysa eğitiminin donanımsal gerçekleşmesi," *Türkiye Bilişim Vakfı Bilgisayar Bilimleri ve Mühendisliği Dergisi*, vol. 5, no. 5, pp. 1-7, 2012.
- [61] B.M. Wilamowski, Y. Chen, and A. Malinowski, "Efficient algorithm for training neural networks

with one hidden layer," *Proceedings of the International Joint Conference on Neural Networks*, vol. 3, 1999. pp. 1725-1728.

[62] J. Arif, N.R. Chaudhuri, S. Ray, and B. Chaudhuri, "Online levenberg-marquardt algorithm for neural network based estimation and control of power systems," *Proceedings of the International Joint Conference on Neural Networks*, 2009, pp. 199-206.

[63] Y. Yang, X.M. Fu and L. Liu, "Computing surface polycube-maps by constrained voxelization," *Computer Graphics Forum*, vol. 30, no. 7, pp. 299-309, 2019.

[64] L. Feng and S.H. Soon, "An effective 3d seed fill algorithm," *Computers & Graphics*, vol. 22, no. 5, pp. 641-644, 1998.

[65] W. E. Lorensen and H.E. Cline, "History of the marching cubes algorithm," *Computer Graphics*, vol. 21, no. 4, pp. 163-169, 1987.

[66] W.E. Lorensen, "Marching cubes: a high-resolution 3d surface construction algorithm," *IEEE Computer Graphics and Applications*, vol. 40, no. 2, pp. 8-15, 2020.

[67] Y. Liao, S. Donne, and A. Geiger, "Deep marching cubes: learning explicit surface representations," *2018 IEEE/CVF Conference on Computer Vision and Pattern Recognition*, 2018, pp. 2916-2925.

[68] K. Ayar, G. Cit, C. Oz and S. Serttas, "Voxelization with OpenCL for virtual sculpting," *International Symposium on Innovative Technologies in Engineering and Science*, 2014, pp. 826-831.

[69] A. Ozkurt, "Surface model extraction from three dimensional sampled data," *Dokuz Eylül Üniversitesi Mühendislik Fakültesi Fen ve Mühendislik Dergisi*, vol. 4, no. 3, pp. 27-36, 2002.

[70] J. Han, "MRI and CT image based on 3d reconstruction and medical rapid prototyping," M.S. Thesis, University of Puerto Rico, Puerto Rico, 2005.

[71] A.N. Chernikov and J. Xu, "Proof of correctness of a marching cubes algorithm carried out with coq," *Proceedings of the 22nd International Meshing Roundtable*, 2014, pp.505-523.

[72] B.N. Parmar and T. Bhatt, "Volume visualization using marching cubes algorithms : survey & analysis," *International Journal of Innovative Research in Technology*, vol. 2, no. 11, pp. 21-25, 2016.

[73] S. Roy and P. Augustine, "Comparative study of marching cubes algorithms for the conversion of 2d image to 3d," *International Journal of Computational Intelligence Research*, vol. 13, no. 3, pp. 327-337, 2017.