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# Prediction of the Welding Process Parameters and the Weld Bead Geometry for Robotic Welding Applications with Adaptive Neuro-Fuzzy Models

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## Abstract

The weld bead geometry is the important information for determining the quality and mechanical properties of the weldment. The welding process parameters or variables that affect the weld bead geometry in the conventional arc welding process include the following: the welding voltage  $U$ , the welding current  $I$ , the wire feed speed  $WFS$ , the contact tip to work distance  $D$ , and the welding speed  $S$ . Modeling and predicting the weld bead geometry play an important role in welding process planning, to determine the optimal welding process parameters for achieving the improved weld quality. There have been lots of efforts and studies to develop modeling solutions and simulations to determine the weld bead geometry (Height  $H$  and Width  $W$ ) from the welding process parameters ( $U$ ,  $I$ ,  $WFS$ ,  $D$ ,  $S$ ) as the inputs. The welding process parameters can be determined based on the experiences, and the conventional analysis of variance (ANOVA); however, the high welding quality and accuracy are not always obtained. With the advancement of computer vision technologies, digital images from cameras and videos can be used for training the deep learning models, to accurately identify and classify objects. The digital images for evaluating the welding quality and the characteristics of welding objects can be captured via the use of the high-speed

camera, and there are emerging data acquisition systems that can handle a huge dataset. In this paper, an adaptive neuro-fuzzy inference system (ANFIS) model is proposed to determine weld bead geometry from the main welding process parameters U, I and S. The proposed ANFIS model was successfully developed for the first basic investigations, as the foundation for further developments of innovative robotic welding systems which can be used for higher educations or research in Smart Manufacturing, with potentials for industrial applications.

## Keywords

- **Welding robots**
- **ANOVA**
- **ANFIS**
- **weld bead geometry**
- **GMA welding**
- **gas metal arc welding**
- **GMAW**
- **metal inert gas welding**
- **MIG**

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## References

1. Petrik, J., et al.: Beyond parabolic weld bead models: AI-based 3D reconstruction of weld beads under transient conditions in wire-arc additive manufacturing. *J. Mater. Process. Technol.* **302**(2022), 117457 (2022)

---

[CrossRef](#) [Google Scholar](#)

2. Le, C.H., et al.: Challenges and conceptual framework to develop heavy-load manipulators for smart factories. *Int. J. Mechatron. Appl. Mech.* **2020**(8), 209–216 (2020)

---

[Google Scholar](#)

3. Arey, D., et al.: Lean industry 4.0: a digital value stream approach to process improvement. *Procedia Manuf.* **54**, 19–24 (2021). ISSN 2351-9789

---

[Google Scholar](#)

4. Daniel, A., et al.: An investigation into the adoption of automation in the aerospace manufacturing industry. In: *Advances in Manufacturing Technology XXXIII*, pp. 87–92. IOS Press (2019). ISBN: 1643680099. <https://doi.org/10.1007/s00170-018-1897-x>
5. Nguyen, T.-T., Le, C.-H.: Optimization of compressed air assisted-turning-burnishing process for improving machining quality, energy reduction and cost-effectiveness. *J. Eng. Manuf. Proc. Inst. Mech. Eng. Part B* **235**(6–7) (2020). <https://doi.org/10.1177/0954405420976661>

6. Singh, C.J., et al.: Optimization of FFF process parameters by naked mole-rat algorithms with enhanced exploration and exploitation capabilities. *Polymers* **13**(11), 1702, 2073–4360 (2021). <https://doi.org/10.3390/polym13111702>
7. Yu, Z., He, Y., Xu, Y., Chen, H.: Vision-based deviation extraction for three-dimensional control in robotic welding with steel sheet. *Int. J. Adv. Manuf. Technol.* **95**(9–12), 4449–4458 (2018). <https://doi.org/10.1007/s00170-017-1546-9>

---

[CrossRef](#) [Google Scholar](#)

8. Wang, X.: Three-dimensional vision-based sensing of GTAW: a review. *Int. J. Adv. Manuf. Technol.* **72**(1–4), 333–345 (2014). <https://doi.org/10.1007/s00170-014-5659-0>

---

[CrossRef](#) [Google Scholar](#)

9. Wang, Z.: An imaging and measurement system for robust reconstruction of weld pool during arc welding. *IEEE Trans. Ind. Electron.* **62**(8), 5109–5118 (2015). <https://doi.org/10.1109/TIE.2015.2405494>

---

[CrossRef](#) [Google Scholar](#)

10. Wang, X.: Three-dimensional vision applications in GTAW process modeling and control. *Int. J. Adv. Manuf. Technol.* **80**(9–12), 1601–1611 (2015). <https://doi.org/10.1007/s00170-015-7063-9>

---

[CrossRef](#) [Google Scholar](#)

11. Chen, H., et al.: Closed-loop control of robotic arc welding system with full-penetration monitoring. *J. Intell. Robot. Syst.* **56**, Article no. 565 (2009)

---

[Google Scholar](#)

12. Thao, D.T., Kim, I.S., Na, H.H., Jung, S.M., Shim, J.Y.: Development of mathematical model with a genetic algorithm for automatic GMA welding process. *Int. J. Adv. Manuf. Technol.* **73**(5–8), 837–847 (2014). <https://doi.org/10.1007/s00170-014-5842-3>

---

[CrossRef](#) [Google Scholar](#)

13. Aviles-Viñas, J.F., Rios-Cabrera, R., Lopez-Juarez, I.: On-line learning of welding bead geometry in industrial robots. *Int. J. Adv. Manuf. Technol.* **83**(1–4), 217–231 (2015). <https://doi.org/10.1007/s00170-015-7422-6>

---

[CrossRef](#) [Google Scholar](#)

14. Keshmiri, S., et al.: Application of deep neural network in estimation of the weld bead parameters. In: 2015 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS) (2015). <https://doi.org/10.1109/IROS.2015.7353868>
15. Vu, M.D., et al.: A conceptual digital twin for cost-effective development of a welding robotic system for smart manufacturing. In: Long, B.T., Kim, Y.-H., Ishizaki,

K., Toan, N.D., Parinov, I.A., Vu, N.P. (eds.) MMMS 2020. LNME, pp. 1018–1025. Springer, Cham (2021). [https://doi.org/10.1007/978-3-030-69610-8\\_134](https://doi.org/10.1007/978-3-030-69610-8_134)

---

[CrossRef](#) [Google Scholar](#)

- 
16. My, C.A., et al.: Inverse kinematic control algorithm for a welding robot-positioner system to trace a 3D complex curve. In: 2019 International Conference on Advanced Technologies for Communications (ATC). IEEE (2019). <https://doi.org/10.1109/ATC.2019.8924540>

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