

REFERENCES

- Farahani, A., Voghoei, S., Rasheed, K., & Arabnia, H. R. (2020). A brief review of domain adaptation. *arXiv preprint arXiv:2010.03978*.
<https://doi.org/10.48550/arXiv.2010.03978>
- Ganin, Y., & Lempitsky, V. (2015). Unsupervised domain adaptation by backpropagation. In *Proceedings of the 32nd International Conference on Machine Learning* (pp. 1180-1189). PMLR.
<https://proceedings.mlr.press/v37/ganin15.html>
- Hasani, M., & Khotanlou, H. (2019). An empirical study on position of the batch normalization layer in convolutional neural networks. In *2019 5th Iranian Conference on Signal Processing and Intelligent Systems (ICSPIS)* (pp. 1-4). IEEE. <https://doi.org/10.1109/ICSPIS48872.2019.9066113>
- He, K., Zhang, X., Ren, S., & Sun, J. (2015). Deep residual learning for image recognition. *arXiv preprint arXiv:1512.03385*.
<https://doi.org/10.48550/arXiv.1512.03385>
- International Foundation for Electoral Systems. (2024). *2024 elections FAQ final*.
https://www.ifes.org/sites/default/files/2024-02/2024_Elections_FAQ_Final.pdf
- Ioffe, S., & Szegedy, C. (2015). Batch normalization: Accelerating deep network training by reducing internal covariate shift. *arXiv preprint arXiv:1502.03167*.
<https://doi.org/10.48550/arXiv.1502.03167>
- Jakarta Globe. (2024). Technical issues cause discrepancies in vote counting machine data. Retrieved April 10, 2025, from <https://jakartaglobe.id/news/technical-issues-cause-discrepancies-in-vote-counting-machine-data>
- Javadecompilers.com. (2025). APK decompiler - decompile Android .apk. Retrieved April 13, 2025, from <http://www.javadecompilers.com/apk>
- Kawal Pemilu. (2024). Real time election monitoring and data verification. Retrieved December 6, 2024, from <https://kawalpemilu.org/>
- Koh, P. W., Sagawa, S., Marklund, H., Xie, S. M., Zhang, M., Balsubramani, A., Hu, W., Yasunaga, M., Phillips, R. L., Gao, I., Lee, T., David, E., Stavness, I., Guo, W., Earnshaw, B., Haque, I. S., Beery, S. M., Leskovec, J., Kundaje, A., ... Liang, P. (2021). WILDS: A benchmark of in-the-wild distribution shifts. In *Proceedings*

Komisi Pemilihan Umum Republik Indonesia. (2025). *Pemilu 2024 - Hasil*

perhitungan suara. Retrieved January 10, 2025, from

<https://pemilu2024.kpu.go.id/>

Krause, A., Perona, P., & Gomes, R. (2010). Discriminative clustering by regularized information maximization. In *Advances in Neural Information Processing Systems*. Curran Associates, Inc.

https://papers.nips.cc/paper_files/paper/2010/hash/42998cf32d552343bc8e460416382dca-Abstract.html

Kumari, T., Vardan, Y., Shambharkar, P. G., & Gandhi, Y. (2022). Comparative study on handwritten digit recognition classifier using CNN and machine learning algorithms. In *2022 6th International Conference on Computing Methodologies and Communication (ICCMC)* (pp. 882-888). IEEE.

<https://doi.org/10.1109/ICCMC53470.2022.9753756>

LeCun, Y., Bottou, L., Bengio, Y., & Haffner, P. (1998). Gradient-based learning applied to document recognition. *Proceedings of the IEEE*, 86(11), 2278-2324.

<https://doi.org/10.1109/5.726791>

Liang, J., He, R., & Tan, T. (2023). A comprehensive survey on test-time adaptation under distribution shifts. *arXiv preprint arXiv:2303.15361*.

<https://doi.org/10.48550/arXiv.2303.15361>

Liang, J., Hu, D., & Feng, J. (2020). Do we really need to access the source data? Source hypothesis transfer for unsupervised domain adaptation. In *Proceedings of the 37th International Conference on Machine Learning* (pp. 6028-6039). PMLR.

<https://proceedings.mlr.press/v119/liang20a.html>

Liu, C.-L., Nakashima, K., Sako, H., & Fujisawa, H. (2003). Handwritten digit recognition: Benchmarking of state-of-the-art techniques. *Pattern Recognition*, 36(10), 2271-2285. [https://doi.org/10.1016/S0031-3203\(03\)00085-2](https://doi.org/10.1016/S0031-3203(03)00085-2)

Mahkamah Konstitusi Republik Indonesia. (2024). IT expert reveals three sources of Sirekap issues. Retrieved April 10, 2025, from

McMahan, H. B., Moore, E., Ramage, D., Hampson, S., & y Arcas, B. A. (2023).

Communication-efficient learning of deep networks from decentralized data.

arXiv preprint arXiv:1602.05629. <https://doi.org/10.48550/arXiv.1602.05629>

Netron. (2025). *Netron*. Retrieved April 13, 2025, from <https://netron.app/>

Niu, S., Wu, J., Zhang, Y., Chen, Y., Zheng, S., Zhao, P., & Tan, M. (2023). Towards stable test-time adaptation in dynamic wild world. In *International Conference on Learning Representations*.

ONNX. (2025). ONNX | Home. Retrieved April 16, 2025, from <https://onnx.ai/>

Papers with Code. (2025a). MNIST benchmark (handwritten digit recognition).

Retrieved April 10, 2025, from <https://paperswithcode.com/sota/handwritten-digit-recognition-on-mnist>

Papers with Code. (2025b). USPS benchmark (image clustering). Retrieved April 10, 2025, from <https://paperswithcode.com/sota/image-clustering-on-usps>

PyTorch. (2025). *PyTorch*. Retrieved April 16, 2025, from <https://pytorch.org/>

Ramazanov, M., Pardo, A., Ghanem, B., & Alfarra, M. (2025). Test-time adaptation for combating missing modalities in egocentric videos. *arXiv preprint arXiv:2404.15161*. <https://doi.org/10.48550/arXiv.2404.15161>

Salma. (2024). Indonesia's 2024 elections: Balancing technology and trust in digital age. *Universitas Gadjah Mada*. Retrieved April 10, 2025, from <https://ugm.ac.id/en/news/indonesias-2024-elections-balancing-technology-and-trust-in-digital-age/>

Simonyan, K., & Zisserman, A. (2015). Very deep convolutional networks for large-scale image recognition. *arXiv preprint arXiv:1409.1556*. <https://doi.org/10.48550/arXiv.1409.1556>

SIREKAP. (2025). Download SIREKAP 2024 2.48 Android APK file.

APKPure.com. Retrieved April 13, 2025, from <https://apkpure.com/sirekap-2024/id.go.kpu.sirekap2024/download/2.48>

Sun, Y., Tzeng, E., Darrell, T., & Efros, A. A. (2019). Unsupervised domain adaptation through self-supervision. *arXiv preprint arXiv:1909.11825*.
<https://doi.org/10.48550/arXiv.1909.11825>

Sun, Y., Wang, X., Liu, Z., Miller, J., Efros, A., & Hardt, M. (2020). Test-time training with self-supervision for generalization under distribution shifts. In *Proceedings of the 37th International Conference on Machine Learning* (pp. 9229-9248). PMLR. <https://proceedings.mlr.press/v119/sun20b.html>

Sun, Y., et al. (2025). Learning to (learn at test time): RNNs with expressive hidden states. *arXiv preprint arXiv:2407.04620*.
<https://doi.org/10.48550/arXiv.2407.04620>

Szegedy, C., Liu, W., Jia, Y., Sermanet, P., Reed, S., Anguelov, D., Erhan, D., Vanhoucke, V., & Rabinovich, A. (2014). Going deeper with convolutions. *arXiv preprint arXiv:1409.4842*. <https://doi.org/10.48550/arXiv.1409.4842>

TensorFlow. (2025). *TensorFlow*. Retrieved April 16, 2025, from <https://www.tensorflow.org/>

tim-learn. (2025). tim-learn/SHOT: Code released for our ICML 2020 paper "Do we really need to access the source data? Source hypothesis transfer for unsupervised domain adaptation". Retrieved April 13, 2025, from <https://github.com/tim-learn/SHOT/>

Wang, D. (2025). *DequanWang/tent* [Computer software]. GitHub.
<https://github.com/DequanWang/tent>

Wang, D., Shelhamer, E., Liu, S., Olshausen, B., & Darrell, T. (2020). Tent: Fully test-time adaptation by entropy minimization. In *International Conference on Learning Representations*. <https://openreview.net/forum?id=uXl3bZLkr3c>

Yohanes. (2025, February 23). Reverse engineering model handwritten digits recognition aplikasi mobile SIREKAP. *Amazing Grace*. Retrieved April 7, 2025, from <https://blog.compactbyte.com/2024/02/23/reverse-engineering-model-handwritten-digits-recognition-aplikasi-mobile-sirekap/>

Zhao, H., Liu, Y., Alahi, A., & Lin, T. (2023). On pitfalls of test-time adaptation. In *Proceedings of the 40th International Conference on Machine Learning* (pp. 42058-42080). PMLR. <https://proceedings.mlr.press/v202/zhao23d.html>

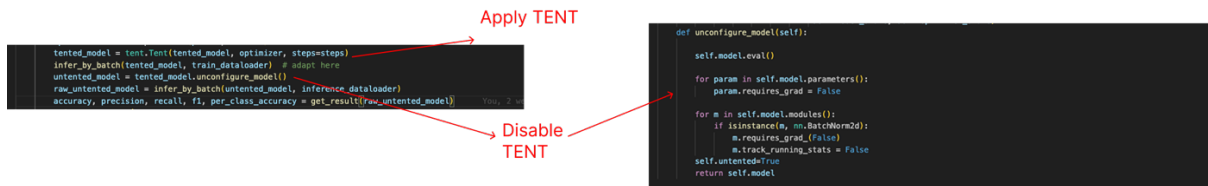


APPENDIX

1. Post-processing ‘applyHeuristic’ code gained from decompilation using javadecompilers.n

```
/* JADK DEBUG: Multi-variable search result rejected for TypeSearchVarInfo(r2v8, resolved type: java.lang.Number) */
/* JADK WARNING: Multi-variable type inference failed */
/* Code decompiled incorrectly, please refer to instructions dump. */
private final int applyHeuristic(float[] r11) {
    /*
        r10 = this;
        kotlin.ranges.IntRange r0 = kotlin.collections.ArraysKt.getIndices((float[]) r11)
        java.lang.Iterable r0 = (java.lang.Iterable) r0
        java.util.Iterator r0 = r0.iterator()
        boolean r1 = r0.hasNext()
        if (r1 != 0) goto L_0x0012
        r0 = 0
        goto L_0x0043
        L_0x0012:
        java.lang.Object r1 = r0.next()
        boolean r2 = r0.hasNext()
        if (r2 != 0) goto L_0x001e
        L_0x001c:
        r0 = r1
        goto L_0x0043
        L_0x001e:
        r2 = r1
        java.lang.Number r2 = (java.lang.Number) r2
        int r2 = r2.intValue()
        r2 = r1[r2]
        L_0x0027:
        java.lang.Object r3 = r0.next()
        r4 = r3
        java.lang.Number r4 = (java.lang.Number) r4
        int r4 = r4.intValue()
        r4 = r1[r4]
        int r5 = java.lang.Float.compare(r2, r4)
        if (r5 == 0) goto L_0x003c
        r1 = r3
        r2 = r4
        L_0x003c:
        boolean r3 = r0.hasNext()
        if (r3 != 0) goto L_0x0027
        goto L_0x001c
        L_0x0043:
        java.lang.Integer r0 = (java.lang.Integer) r0
        r1 = 0
        if (r0 == 0) goto L_0x004d
        int r0 = r0.intValue()
        goto L_0x004e
        L_0x004d:
        r0 = r1
        L_0x004e:
        return r0
    */
}
```

2. Code For TENT Evaluation in Adaptation Size Evaluation



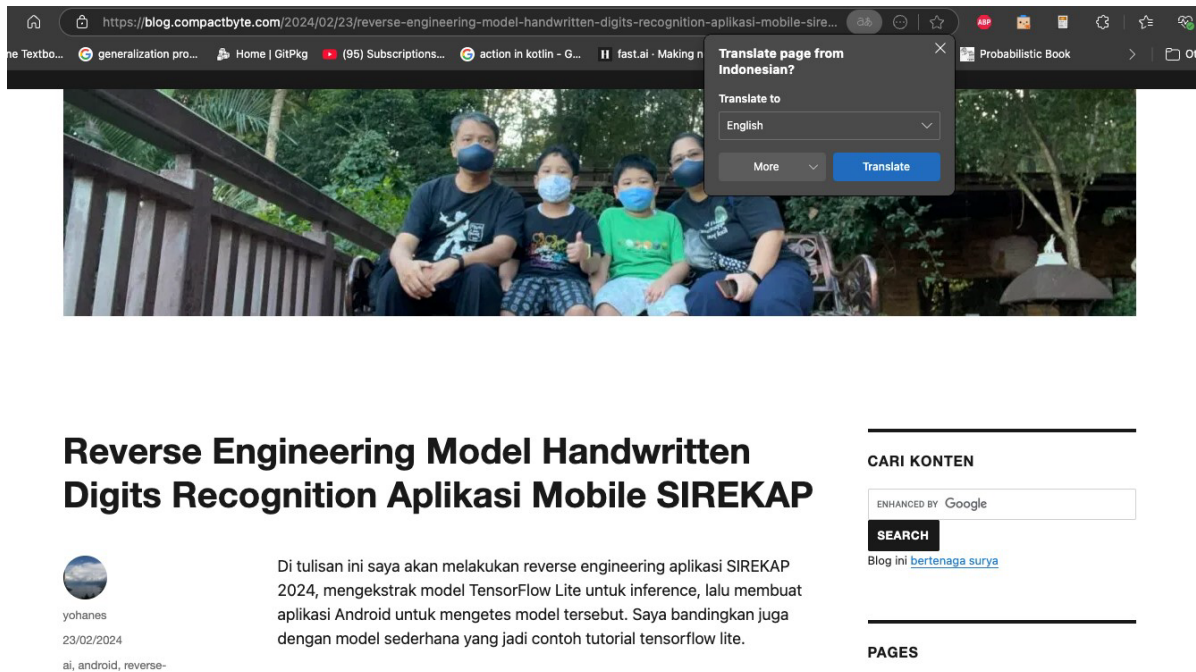
3. Code For SHOT Evaluation in Adaptation Size Evaluation

```

if iter_num % interval_iter == 0 or iter_num == max_iter:
    netF.eval()
    netB.eval()
    acc, _, precision, recall, f1, class_accuracies = cal_acc_key_metrics(dset_loaders['test'], netF, netB, netC) # Real accuracy after running n times
    log_str = "Task: {}, Iter: {}/{}; Accuracy = {:.2f}%".format(args.dset, iter_num, max_iter, acc)
    saveResultTestAccuracyOnce, f1, precision, recall, args.iteration, args.dset_size, iter_num, max_iter, "test_accuracy_with_max_iter_iter_num_key_metrics_tracked", "/workspace/Dwika/fyp/saved_results_with_key_metrics")
    columns = ["Class", "Accuracy", "iter_num", "max_iter"]

    for class_label, accuracy in class_accuracies.items():
        write_row_to_csv(f"/workspace/Dwika/fyp/saved_results_with_key_metrics/per_class_acc/{args.dset_size}_{args.iteration}.csv", columns, [class_label, accuracy, iter_num, max_iter])

print(log_str)
netF.train()
netB.train()
    
```



5. **Github Link: https://github.com/Dwikavindra/Final_Year (with details written in README.MD)**