
Master's thesis on the topic of

**“Continual Learning Method for Image
Classification in Computer Vision”**

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Outline

- Continual Learning
- Parking Lot Occupancy Detection
- Baseline Model
- Reproduction
- Results
- Conclusion

Continual Learning

- For decades, attempts have been made to use AI to automate difficult tasks.
- One feature of human intelligence has still been difficult to reproduce using AI – lifelong learning.
- Continual Learning (CL) is a set of approaches within ML that aims to address this limitation.
- Why can our models not learn from their production experience and improve further?

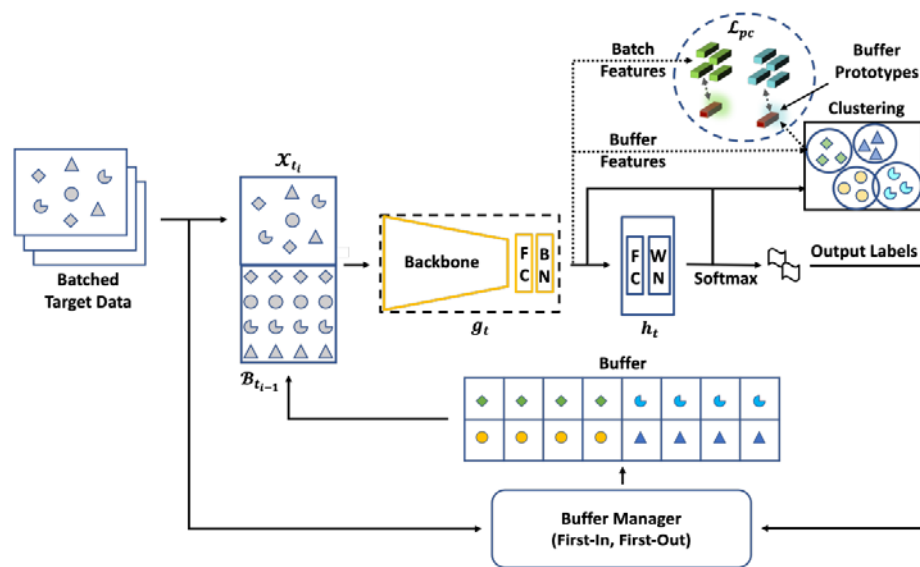
Parking Lot Occupancy Detection

- Binary classification problem, with an endless stream of input data in the form of video camera feed.
- A rather simple problem in the field of ML, but is a perfect use case for CL.
- An amazing comprehensive survey by de Almeida et. al. covers public datasets for this problem, and lists reproducible and unbiased state-of-the-art solutions.

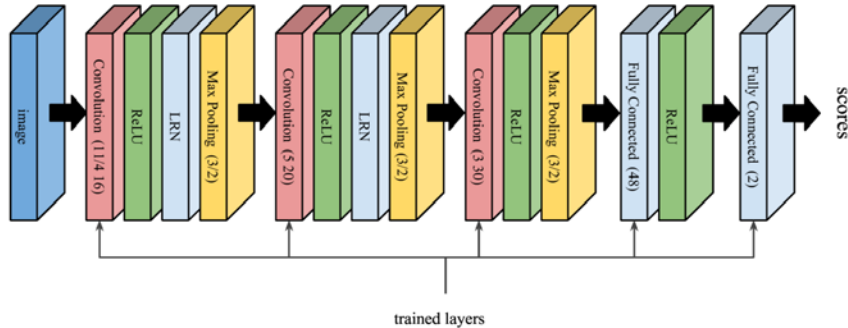


How it can be improved with CL?

- The UCL-GV method was proposed by Taufique et. al. that performs Unsupervised Domain Adaptation in gradually varying domains.
- This can allow us to tackle gradually changing light conditions, weather changes, and more long-term seasonal changes in the input data.



Baseline model



- The baseline model was chosen to be mAlexNet by Amato et. al.
- Very simple, yet achieves excellent results.
- Well documented, with publicly available source code.
- Can be combined with UCL-GV in an attempt to improve its performance in a gradually varying domain scenario.

Reproduction

- After re-implementing the model, experiments have been performed to verify that results can be achieved that are similar to the ones reported in the original paper.
- Most experiments differ from the original by no more than 5%, even outperforming the original model in some cases.

Method	Train DS / split	Test DS / split	Accuracy	AUC
Reproduced	CNRPark / Odd	CNRPark / Even	91.43%	0.97
Original	CNRPark / Odd	CNRPark / Even	90.13%	0.94
Reproduced	CNRPark / Even	CNRPark / Odd	91.5%	0.97
Original	CNRPark / Even	CNRPark / Odd	90.71%	0.92
Reproduced	CNRPark / All	CNRParkEXT / Test	93.64%	0.98
Original	CNRPark / All	CNRParkEXT / Test	93.52%	0.98
Reproduced	CNRPark / All	PKLot / TwoDays	85.08%	0.93
Original	CNRPark / All	PKLot / TwoDays	90.38%	0.99
Reproduced	PKLot / Train	CNRParkEXT / Test	74.77%	0.85
Original	PKLot / Train	CNRParkEXT / Test	83.83%	0.91
Reproduced	PKLot / Train	PKLot / Test	84.29%	0.97
Original	PKLot / Train	PKLot / Test	98.07%	0.99
Reproduced	CNRParkEXT / Cam1	CNRParkEXT / Cam2	95.34%	0.99
Original	CNRParkEXT / Cam1	CNRParkEXT / Cam2	95%	-

Results

- Experiments have been modeled to feed images in the order they were taken.
- Experiment results consistently show that the model that is enhanced with CL techniques is able to consistently achieve better accuracy on a classical ML scenario.

Method	Train DS / split	Test DS / split	Accuracy
mAlexNet	CNRParkEXT / All	CNRPark / All	95.21%
mAlexNet+U CL-GV	CNRParkEXT / All	CNRPark / All	95.73%
mAlexNet	CNRParkEXT / Cam1-8	CNRParkEXT / Cam9	96.01%
mAlexNet+U CL-GV	CNRParkEXT / Cam1-8	CNRParkEXT / Cam9	96.85%
mAlexNet	PKLot / TwoDays	PKLot / NotTwoDays	91.74%
mAlexNet+U CL-GV	PKLot / TwoDays	PKLot / NotTwoDays	92.11%

Conclusion

- The paper attempts to view the paradigm of Continual Learning from a completely new perspective.
- CL methodologies have been successfully utilized to achieve better performance on traditional problems, as opposed to solving new types of problems.
- Even “mostly solved” ML problems can be further improved by introducing contemporary CL methods.
- Future work may include attempts to improve more classical ML problems using this technique.

Thank you for your attention!

