Reinforcement Learning for Bomberman

Final Project for the lecture: Fundamentals of Machine Learning

Daniel Gonzalez, Matrikel Nr.: 3112012 Maria Regina Lily, Matrikel Nr.: :) Ferdinand Vanmaele, Matrikel Nr.: ;) (Order alphebitcally, don't take it personal)

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1 Abstract

SOME TEXT - Look at the en in Conclusion for some useful links



Figure 1: he acts as a dement (left) he is often seen as a terrorist(right)

2 Introduction

[TODO: some text]

3 Explaining the Framework

Features:

- A1: Assumption:
 - 1) BOMB -> others[i] -> agent (don't die)
 - 2) BOMB -> BOMB -> agent (die)
 - 3) BOMB -> crate -> agent (don't die)
- C1: Consideration:

use s.bomb power

C2: Consideration:

use bombs left

- FOR STEP 1:
 - 1.) Reward the best possible action to a coin, if it is reachable F(s,a)=1, otherwise F(s,a)=0. 'BOMB' and 'WAIT' are always 0.
- FOR STEP 1 & 2:
 - -1.)

DONE

- 2.) Penalize if the action follows the agent to death (F,s)=1, F(s,a)=0. otherwise.
 DONE, it could be slightly improved
 - REMARK: C1?
- 3.) Penalize if the action follows the agent into a "save": (Where the bomb won't at some point explode) position. F(s,a)=1 otherwise F(s,a)=0. Bombs are always set to 0. **TO BE DONE**
- **4.)** Reward the minimal distance that follows to safety if you are in a "Danger zone" (as defined in **3**.) F(s,a)=1 otherwise F(s,a)=0. If you are not in a "Danger zone" then F(s,a)=0 for all actions. For Bombs always set F(s,a)=0.

TO BE DONE: -> LILY

- 5.) Penalize invalid actions. F(s,a) = 1, otherwise F(s,a) = 0.

DONE

REMARK: C2

- 6.) Reward when getting a coin $F(s,\!a)=1,$ otherwise $F(s,\!a)=0$. ${\bf DONE}$
- 7.) Reward putting a bomb next to a block.
 F(s,a) = 0 for all actions and otherwise F(s,a) = 1 for a BOMB if we are next to a

block. **DONE**

– 8.) ## Reward the available movements F(s,a)=1, otherwise F(s,a)=0. Remark: TO BE DEFINE and DONE

Weitere Ideen:

• penalize distance between our agent and the nearest reachable crate.

TD(0) learning

[TODO: some text] Probably not useful, but maybe useful as idea for loading images







Some text

Figure 2: Some text

Gradient descent

[TODO: some text] Probably not useful, but maybe useful as idea for loading images



Figure 3: Some text

Bla blah

Other sub chapter

[SOME TEXT].



Figure 4: some text

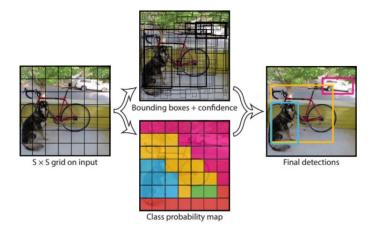


Figure 7: SOME TEXT.

Related work

[Some text]

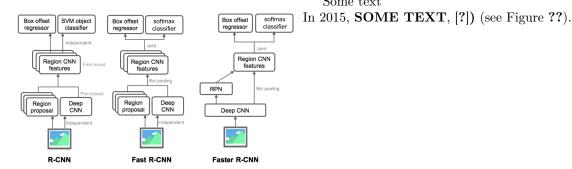


Figure 5: SOME TEXT

bla bla

some citation examples [?][?]

See figures above (Strategy to also use space and have more 5 variants for the text

Some text

SOME TEXT

The YOLO approach to object detection

SOME TEXT

Anchor boxes

SOME TEXT

SUBSECTION

SOME TEXT

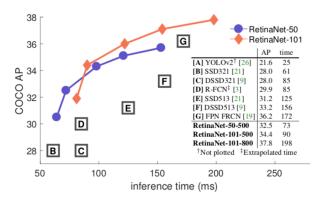


Figure 6: some text

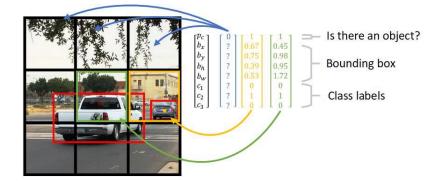


Figure 8: SOME TEXT

SUBSECTION

6 Feature extraction

Some text

7 Evaluation

Some text

8 Results: some Table & other way of loading images

some text

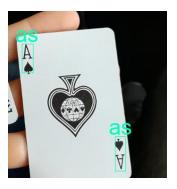
some table

Dataset situation				
name	description	precision	recall	mAP
1 - Simple	Paste cards on simple canvases	0.974	0.996	0.991
	random rotations, brightness, blurring			
2 - Medium	Paste randomly scaled cards on simple canvases	0.946	0.988	0.989
	random rotations, brightness, blurring			
3 - Elaborate	Paste randomly scaled cards on textures	0.937	0.978	0.971
	random rotations, brightness, blurring			
4 - Hardest	Paste randomly scaled cards on textures	0.940	0.983	0.973
	$random\ rotations,\ brightness,\ blurring,\ less\ zoom$			

Table 1: Precision and recall values have been calculated using a IOU threshold of 0.5. mAP values are based on averaged precision values over IOU thresholds of $[0.1, 0.2, \dots 0.8, 0.9]$



success: classification: **Ad:** 0.99995, **Ad:** 0.99997



success: classification **As:** 0.99757, **As:** 0.99931



success: classification **Jd:** 0.99967, **Jd:** 0.99992

Figure 9: Successful cases of detection of images that are pretty representative of the training distribution

Results of further work

Transfer learning

Webcam deployment

9 Discussion and Future Work [Frank & Daniel]

Overview

Training process

Deployment on a webcam

Future Work

10 Conclusion

https://github.com/mlteam-ws2018/RL_boom. SOME USEFUL LINKS (for the reportwriting):

[for motivation]: https://www.youtube.com/watch?v=xMP-JqFQ_14

 $[gd, policy, q-learning]: \verb|https://www.ias.informatik.tu-darmstadt.de/uploads/Theses/Sharma_BScThesis_2012.pdf|$

 $[gd, policy, q-learning]: \verb|https://repositorio-aberto.up.pt/bitstream/10216/91011/2/176444.| pdf$

References

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