Playing a class of Game using CNN Focus on Runner Games

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A presentation for Communicative English course



- 1 Introduction
- 2 Selection of a wrong game!
- 3 Why pong failed?
- 4 Discovering a new Class of Runner Games
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This was a pet project during my 2nd year on St. Xavier's College!

Motivation





Figure: Indranil and Harrison



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Figure: Indranil and Harrison

- CNN to analyze visual imagery, face detection, earlier application include detection of numbers for post cards (Yann Lee Cunn)
- Cifar 10 model, 80% accuracy in 2010 on CIFAR-10 dataset
- Can we make something innovative using simple technology?

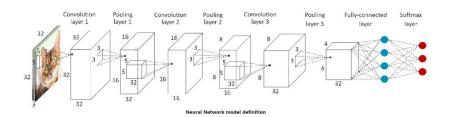


Figure: The famous CIFAR-10 model which we used for training



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Selection of a wrong game!

The game that came to our mind - PONG! Pong - Simplest table tennis video game from 1972 atari console.

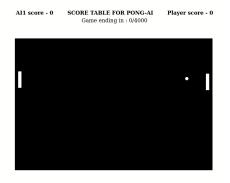


Figure: Uprighted Cabinet of Pong

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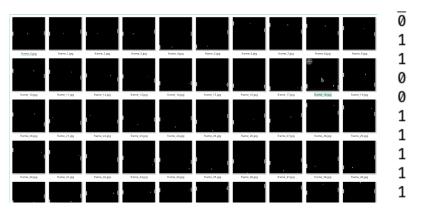
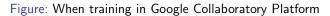


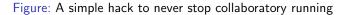
Figure: Data used in training of Pong Game: Frames and corresponding action values (0 - down and 1 - up)

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Epoch 22/50
20000/20000 [========== ] - 209s
Epoch 23/50
☐ Inspector ☐ Console ☐ Debugger ↑ Network {} Style
   Filter Output
>> allow pasting
SyntaxError: unexpected token: identifier [Learn More]
>> function ClickConnect(){
  console.log("Working");
  document.querySelector("colab-toolbar-button#connect").click()
  setInterval(ClickConnect,60000)
← 16465
  Working
  Working
```





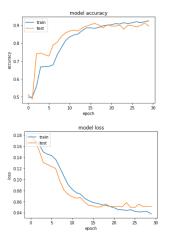


Figure: Loss and accuracy of the Pong game trained with 2K images

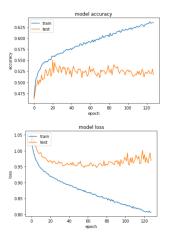


Figure: Loss and accuracy of the Pong game when trained with 30K images



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Pong's Computer counterpart is a robot not an AI, it just calculates according to coordinates of the ball. We needed human touch!

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- More data more accuracy, what is wrong?
- We trained the model using Google's free collaboratory platform, which gives free GPU and is computationally effective.
- Guess the move by looking at the Pong's picture



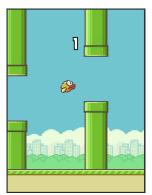
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Figure: Famous Dino game!

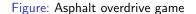
















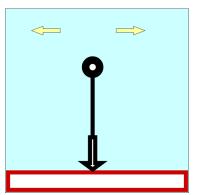


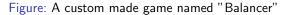














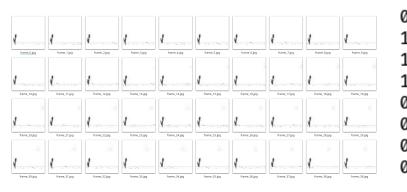


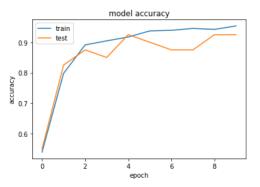
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Results for the Dino game

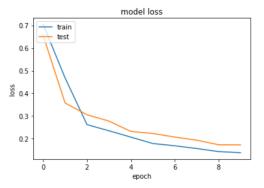
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Conclusion

Modern technologies and algorithms like Reccurrent Neural Network, Reinforcement Learning, Genetic algorithms are more powerful than the method that was implemented. We will implement these in the Dino game in the near future. The YOLO works on this exact same model, i.e., extract frames from the video and predict from those taken pictures.



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Acknowledgements

I acknowledge the help recieved from Tamal Maharaj, Prof. Janardan Ghosh, Indranil Das and Harrison Kinseley for their suggestions and discussions.



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- Pal, J.B., (2019), Playing a class of games using CNN, Blog in github pages, https://jimut123.github.io/blogs/cnn_games_ai.html available on the web, last accessed on 14-11-2019.
- Kinseley, H., (2017), Python Plays GTA V, Tutorial in https://pythonprogramming.net/game-frames-open-cvpython-plays-gta-v/ available on the web, last accessed on 14-11-2019.



Thank You

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