

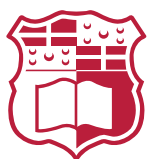
PROGRESS REPORT: Skateboard Trick Recognition through an AI-based Approach

Kris Saliba

Supervisor: Prof. Joseph Bonello

June 2024

*Submitted in partial fulfilment of the requirements
for the degree of Bachelor of Science in Information Technology (Honours)
(Software Development).*



L-Università ta' Malta
Faculty of Information &
Communication Technology

Contents

Contents	i
Abstract	ii
0.1 Brief and represented literature review	2
0.1.1 Accelerometry approach	2
0.2 Planned methodologies	2
0.3 Problem Specification	2
0.4 Project timeline	3

Abstract

This is a test

Introduction

Skateboarding dates back to the 1940s when handmade skateboards first appeared [1]. It has since developed into a worldwide phenomenon, with its popularity skyrocketing, after gaining recognition as an official sport in the 2020 Tokyo Olympics [2]. Skateboarding comprises the dynamic activities of riding a skateboard and skillfully performing a repertoire of tricks, manifesting as a popular and exhilarating “extreme sport”.

This dynamic sport encompasses various disciplines and styles of riding, each of them offering unique challenges for skateboarders to explore. Two of the most prominent styles are “vert” and “Street”, vert skateboarding revolves around riding on specialised obstacles, namely, half-pipes and ramps, emphasising areal manoeuvres. Whilst, street skateboarding transpires in urban environments, utilising a diverse array of obstacles that can be found outdoors, including stairs, rails, ledges, gaps or flat ground for skaters to showcase their creativity. [3].

Skateboard tricks are the heart and soul of skateboarding. These tricks originate from the dynamic orchestration of rotations and revolutions of a skateboard along various axes emphasising the significance of precise placement of a skateboarder’s feet to initiate these rotations. These tricks serve as excellent examples of how the skateboarder’s body and skateboard work in perfect harmony. Some common skateboard tricks include:

- **Ollie:** One of the first tricks beginners learn. Where the skateboarder pops the tail of the board while sliding their foot across the board, causing the board to level out in the air, used to jump over obstacles.
- **Kickflip:** A trick where the skateboarder flips the board under their feet while jumping, making it spin 360° around the x-axis.
- **360 kickflip:** A combination of a kickflip and a 360° board rotation around the y-axis.

Skateboarders continually innovate and come up with new trick combinations, contributing to the dynamic nature of the sport.

General Aims and Goals

The primary goal of this project is to design and develop a dependable Artificial Intelligence (AI) model that can identify various skateboard tricks directly from video. An initial focus will be placed on the recognition three different tricks namely, kickflips, ollies and shuvits. To reach this goal the appropriate ML techniques must be utilised to overcome basic computer vision challenges, such as varying camera angles, lighting conditions and the fast-paced nature of skateboard tricks. By addressing these challenges, this project aims to establish a usable tool that skateboarders, coaches, and skateboard competitions can use.

0.1 Brief and represented literature review

There has been very little research in recent years on the development of a skateboard trick classifier using Artificial Intelligence (AI). In this emergent field, two primary techniques have fostered prominence among researchers. The first technique involves utilising signals obtained from skateboard-mounted accelerometers or artificially generated based on findings of prior studies. These signals are then fed into a study-dependent model for classification, as outlined in [4] and [5]. The second approach employs computer vision techniques, leveraging video footage of skateboard tricks to train and refine models for accurate trick identification, as depicted by the studies [6] and [7].

0.1.1 Accelerometry approach

0.2 Planned methodologies

0.3 Problem Specification

The current lack of a digital and objective mechanism for identifying skateboard tricks during competitions and practice sessions is a significant issue. Judges in live-streamed skateboarding competitions currently announce tricks verbally, without the support of a digital overlay to display the performed trick for the viewers. This reliance on subjective judgement can lead to scoring disparities and conflicts, weakening competition fairness. Furthermore, the absence of an objective method obstructs skaters' ability to receive real-time feedback, which could be beneficial for skill development. The advent of an AI-based system that delivers a dependable and

efficient method of trick recognition will both revolutionise this aspect and contribute to the lack of study in this area.

0.4 Project timeline

Bibliography

- [1] *Encyclopaedia britannica*, 2023. [Online]. Available: <https://www.britannica.com/sports/skateboarding>.
- [2] *One year on: How skateboarding's olympic debut changed the games 2022*, 2022. [Online]. Available: <https://olympics.com/en/news/one-year-on-skateboarding-olympic-debut-feature>.
- [3] Z. Foley, *Vert, street, park - what are the different styles of skateboarding?* 2021. [Online]. Available: <https://www.goskate.com/top/skateboarding-styles-full-guide/>.
- [4] M. A. Abdullah *et al.*, "The classification of skateboarding tricks via transfer learning pipelines," *PeerJ Computer Science*, vol. 7, e680, 2021.
- [5] N. K. Corrêa, J. C. M. d. Lima, T. Russomano, and M. A. d. Santos, "Development of a skateboarding trick classifier using accelerometry and machine learning," *Research on Biomedical Engineering*, vol. 33, pp. 362–369, 2017.
- [6] M. Shapiee *et al.*, "The classification of skateboarding tricks: A transfer learning and machine learning approach," *MEKATRONIKA*, vol. 2, pp. 1–12, Oct. 2020. DOI: 10.15282/mekatronika.v2i2.6683.
- [7] H. Chen, "Skateboardai: The coolest video action recognition for skateboarding (student abstract)," *Proceedings of the AAAI Conference on Artificial Intelligence*, vol. 37, pp. 16 184–16 185, Jun. 2023. DOI: 10.1609/aaai.v37i13.26952.