

# Generative Model of Human Body Motion for Walking

Research Project  
KIREPRO1PE

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## 1 Introduction

short intro to the project

## 2 Background

a chance to write a bit about what i learned from my lit study

## 3 Method

### 3.1 Models

detailed description of my models what makes them unique how they work, etc

### 3.2 Training

notes on how the training was performed

## 4 Results and analysis

a section presenting my results and comparing them to other results, might have to create some "artificial" results to get anything good to compare with, like using linear interpolation as a baseline, and final ik as "competing" solution, this is also a good place to show some concrete examples of what my models can do, using images

### 4.1 Further research and improvements

as with all papers that are things that i did not get to explore, this will detail where i think further energy should be exerted based on my findings.

## 5 Conclusion

## References

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