

# The title of my project

A short description of your project, 3–5 sentences. Expand the description from your previous step.

## 1 Introduction

Table 1 shows a list of solutions to compare. Put here what the table presents, briefly.

Table 1: Comparative analysis of basic solution to my problem, change the title of the table.

Solution	Strengths	Weakness
High-order canonical correlation analysis [1]	Large feasible set of ranking functions; fast convergence to a local optimum	Does not provide global optimum in the feasible set of functions

The section references contain the full list, collected for this project.

## References

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