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Admission Research Proposal Template

For Higher Degree by Research Applicants

Structure your Research Proposal based on the headings provided below, use a clear and legible font (e.g. Arial, size 12) and observe the page/word limit:

A 2,500 word limit (5 pages) applies, excluding Bibliography/ References. 1000 words are normally sufficient.

Name	
Empl ID (if applicable)	
Proposed School/Discipline	
Word Count	

1. Project (research topic) title

Provide a short descriptive title of no more than 20 words.

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2. Project (research topic) summary

In no more than 100 words, intelligible to a reader who is not a specialist in this field, summarize the aims, significance and expected outcomes of your proposed research.

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3. Project (research topic) details

Provide a brief but comprehensive statement, following the headings provided below. Please number the pages. You are urged to avoid the use of jargon and to express your arguments clearly and concisely.

3.1 Introductory background

3.2 Research questions

This section should explain the research question and may include the hypothesis or problem to be addressed.

3.3 Aims/Objectives of the project

This section should explain how the problem is to be addressed.

3.4 Significance/Contribution to the discipline

This section will normally justify the project from a review of the literature on the topic.

3.5 Theoretical framework and methods

This section may include a discussion of the sources to be consulted.

4 Research plan and timeline and contents of thesis/submission

Please provide a plan of stages of your research development and thesis writing with a time-line of proposed goals for your program. A [Gantt chart or Completion Plan](#) is useful in illustrating the various stages and activities of research, in terms of a schedule to meet time-lines.

5 References cited/Bibliography of planned reading

Please provide one or both of the following in an appropriate and consistent referencing style:

- A list of the scholarly references cited in the proposal
- A bibliography of planned reading

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Further Information

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BIBLIOGRAPHY / PLANNED READING

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