# Academic BRASS

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## Good Library Data Made Better With Technology! Using OpenRefine and Google Fusion Tables in Academic Business Libraries Instruction

## Introduction

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## Setting

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## Google Fusion Tables

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Figure 1 Google Fusion Tables example data. I pulled the data from Wikipedia and LexisNexis and limited by temperature and country.

## OpenRefine

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## Putting it all Together in a One-Shot Session

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## Conclusion

This so trancle has covered to bols, Google Fison Tables and OpenRefine, and heir or in bises ordergradate library into to n Google Fison Table offers collaborate and data merging feates in a familiar environmento telent s OpenRefine is a poverflue tismple data cleaning option. These to olshighlight augitibraries can provide sale to heir data ests for no ice big-data analysin aughatamptify the taditonal no le of accessand presentation.

## Works Cited:

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