

1 **TITLE GOES HERE**

2 **Running title:** Optimal resolution

3 Courtney R. Armour, (ande? begum? etc?), Patrick D. Schloss [†]

4 [†] To whom correspondence should be addressed:

5 pschloss@umich.edu

6 Department of Microbiology

7 University of Michigan

8 Ann Arbor, MI 48109

9 **(observation format - max 1200 words, 2 figures, 25 ref)**

¹⁰ **Abstract (max 250 words)**

¹¹ **Importance (max 150 words)**

12 Introduction

- 13 • CRC is one of the most common cancers and a leading cause of cancer related death
- 14 • There is evidence that the microbiome has a role in CRC development/progression and could be useful
15 for biomarker detection and diagnostics.
- 16 • Begum et al (mBio 2020) recently demonstrated effective application of machine learning (ML) to
17 microbiome based classification problems and developed a framework for applying ML practices in a
18 more reproducible way (mikropml).
- 19 • A common question when applying ML methods to microbiome data is which method and taxonomic
20 level should be use.
- 21 • This analysis utilizes the reproducible framework developed by Begum et al to quantify which ML
22 method and taxonomic level produce the best performing classifier for CRC data.

Results

- Of the five ML methods tested, Random forest was consistently the top performer (supplemental figure of all models?) at most taxonomic levels.
 - RF might be more appropriate anyways since its more suitable for zero inflated data? (need to look into literature)
- Within the RF model, the highest AUCs were observed for family, genus, and otu level data with no significant difference between the three. (Figure 1)

31 **Materials and Methods**

- 32 • data from prior study {baxter}
- 33 • mikropml package
- 34 • pvalues as previously described {begum}

