Individual Project Proposal

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Overview of Goals

My graduate research focuses on how soil health and soil microbiology influence long-term sustainability in semiarid wheat systems. Wheat is one of the most important crops worldwide, yet the semiarid landscapes where it is primarily grown are particularly vulnerable to climate change and land degradation (Asseng et al. 2015; USGCRP 2018). In particular, the traditional wheat-fallow system practiced throughout the High Plains ecoregion inefficiently stores water, depletes soil fertility, and has a high potential for erosion, pushing farmers and researchers to search for economical ways to build long-term soil health (Norton, Mukhwana, and Norton 2012; Kaur et al. 2015).

Soils with an active microbiome are critical for sustainable and resilient food production. Agriculture impacts microbial activity through practices such as crop rotation, fallowing, tillage, and fertilization, though these effects can vary depending on environmental factors such as climate and soil type (Lehman et al. 2015). Soil microorganisms are sensitive indicators of changing soil health, and can help evaluate the long-term sustainability of land management practices (Rodgers, Norton, and Diepen 2021). However, there are gaps in our understanding of how chemical, physical, and biological soil properties influence soil microbiology, as well as to how soil microbial properties should be interpreted, especially in semiarid systems. Specifically, this project will use data collected in semiarid grain systems in the Wyoming High Plains to look for correlations between soil microbiology and other, well-established soil health indicators in order to better understand the relationship between soil microbiology and agricultural sustainability in these systems.

Overview of Datasets

I plan to use datasets from my own research on soil health in semiarid wheat systems. These datasets include soil health and management data on plots from several agricultural systems. This includes data on soil physical properties (such as aggregate stability abd bulk density), soil chemical properties (such as total organic matter, labile carbon and nitrogen pools, and pH) and soil microbiology (enzyme activity, microbial biomass, and microbial community composition). This data is mainly in a tidy format, where each row is a plot and each column is a variable describing some soil property. I hope to merge data from different projects, then separate that dataset into different relational databases dealing with different aspects of the plot (sampling and site information, soil physical and chemical properties, and soil microbiology). In analyzing this data, I forsee challenges in dealing with missing values, making sure the keys match exactly between datasets, and ultimately selecting which data and plots to include in my analysis.

```
#Here is one dataset that I plan to use
soil_data <- readxl::read_xlsx('soil_data_OREI.xlsx')
head(soil_data)</pre>
```

```
## 2 A12
                                     NA 2020 comp~ Bare
                                                               fallow
                                                                                    NA
                         NA
                                                                            NA
## 3 B11
                         30
                                     NA 2016 comp~ CC
                                                                                    NA
                                                               fallow
                                                                            NA
## 4 B12
                         45
                                     NA 2016 comp~ Bare
                                                               fallow
                                                                            NA
                                                                                    NA
## 5 C11
                                     NA 2016 comp~ CC
                                                               fallow
                                                                                    NA
                         15
                                                                            NA
## 6 C12
                         30
                                     NA 2016 comp~ Bare
                                                               fallow
                                                                            NA
                                                                                    NA
     ... with 33 more variables: Heads <dbl>, H2O <dbl>, WFPS <dbl>, NO3 <dbl>,
##
       PMN <dbl>, Porosity <dbl>, BulkDensity <dbl>, Protein <dbl>, POXC <dbl>,
       DOC <dbl>, DON <dbl>, PMC <dbl>, MBC_fumigated <dbl>, MBN_fumigated <dbl>,
## #
## #
       BG <dbl>, CBH <dbl>, PHOS <dbl>, NAG <dbl>, BX <dbl>, AG <dbl>, SUL <dbl>,
## #
       LAP <dbl>, InorganicC <dbl>, SOC <dbl>, N <dbl>, actinomycetes <dbl>,
## #
       gram_neg <dbl>, gram_pos <dbl>, AMF <dbl>, sapro_fungi <dbl>,
## #
       MB_PLFA <dbl>, total_fungi <dbl>, total_bacteria <dbl>
```

Proposed Workflow

These datasets will need to be loaded into R and tidied (using tidyverse packages), and then visualized (using ggplot2 and plotly). Specifically, I will be looking for correlations between microbial data and labile carbon/nitrogen pools, which indicate soil health, and displaying those relationships. First, I will use functions from tidyverse, especially merge, mutate, and select. I will merge datasets from different projects, then split that dataframe into relational databases dealing with different aspects of the data, to make them easier to work with. I will use the mutate function to calculate new variables, such as the ratio of fungi to bacteria or labile to total carbon. I will first analyze my dataset using the corr package, which explores correlations between variables in large datasets. Once I find some interesting correlations between microbial properties and soil health variables, I will select my variables of interest. Lastly, I will visualize those relationships using interactive scatterplots from plotly and save them using patchwork and htmlwidgets.

References

Asseng, Senthold, Frank Ewert, Pierre Martre, Reimund P Rötter, David B Lobell, Davide Cammarano, Bruce A Kimball, et al. 2015. "Rising Temperatures Reduce Global Wheat Production." *Nature Climate Change* 5 (2): 143–47.

Kaur, Gurpreet, Axel Garcia y Garcia, Urszula Norton, Tomas Persson, and Thijs Kelleners. 2015. "Effects of Cropping Practices on Water-Use and Water Productivity of Dryland Winter Wheat in the High Plains Ecoregion of Wyoming." *Journal of Crop Improvement* 29 (5): 491–517.

Lehman, R Michael, Veronica Acosta-Martinez, Jeffrey S Buyer, Cynthia A Cambardella, Harold P Collins, Thomas F Ducey, Jonathan J Halvorson, et al. 2015. "Soil Biology for Resilient, Healthy Soil." *Journal of Soil and Water Conservation* 70 (1): 12A–18A.

Norton, Jay B, Eusebius J Mukhwana, and Urszula Norton. 2012. "Loss and Recovery of Soil Organic Carbon and Nitrogen in a Semiarid Agroecosystem." Soil Science Society of America Journal 76 (2): 505–14.

Rodgers, Hannah R, Jay B Norton, and Linda TA van Diepen. 2021. "Effects of Semiarid Wheat Agriculture Management Practices on Soil Microbial Properties: A Review." Agronomy 11 (5): 852. USGCRP, Climate. 2018. "Fourth National Climate Assessment."