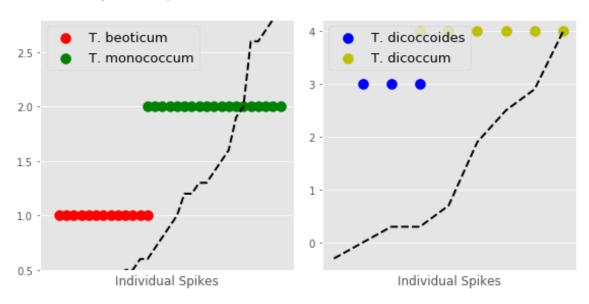
1 Setup

2 Load Model

2.1 Predict together - part 1



2.2 Predict together - part 2

2.2.1 Predict DF Function

```
def make df(model='dom ~ length * depth * width -1 ', model please=False, model test=False, mono=False, di=False
       def aggregate_average_attribute(df, att):
          return df.groupby(['Sample name', 'Sample Type', 'Wild/Domesticated', 'Ploidy'],
                        as_index=False)[att].mean()
       atts = ['length', 'width', 'depth']
       #df = aggregate_average_attribute(pd.concat([einkorn, emmer]), atts)
       if mono:
          df = (pd.concat([einkorn]))
       elif di:
10
          df = (pd.concat([emmer]))
11
       else:
12
          df = (pd.concat([einkorn, emmer]))
13
15
       df = df.sort values(by='Ploidy')
16
       def allocate_ploidy_dom(x):
17
          if (x['Ploidy'] == '2n'):
18
             if (x['Wild/Domesticated'] == 'wild'):
19
                return 1
20
             return 2
21
          else:
22
             if (x['Wild/Domesticated'] == 'wild'):
23
                return 3
24
             return 4
25
26
       df['dom'] = df.apply(allocate ploidy dom, axis=1)
27
       df = df.sort\_values(by='dom')
28
29
       from sklearn.cross_validation import train_test_split
30
       X train, X test, y train, y test = train test split(df[atts], df['dom'], test size=0.2, random state=1)
31
```

```
32
       X_{train}[dom'] = y_{train}
33
       import statsmodels.formula.api as smf
34
       import statsmodels.api as sm
35
36
       if model_test:
37
          model = smf.ols(model, data=X_train).fit()
38
           #print(model.summary())
39
       else:
40
          model = smf.ols(model, data=df).fit()
41
42
       y = df['dom']
43
       x = df[atts]
44
       df['ypred'] = np.around(model.predict(x),2)
46
47
       if model_please:
48
          return model
49
       return df
50
```

Test Model

```
def make test dfs(dom=False):
       a = make df(model test=True, mono=True)
       c = make_df(model_test=True, di=True)
       if dom:
           b=a
           d=c
       else:
          b = a[['Sample Type','ypred']].melt( 'Sample Type', var name='value', value name='res')
          d = c[['Sample Type','ypred']].melt( 'Sample Type', var name='value', value name='res')
10
       def make correction mono(x):
12
          if 'mono' in x['Sample Type']:
13
             if x['res'] != 1:
14
                 return 'Correct'
15
             else:
16
                return 'Incorrect'
17
          if x['res'] != 2:
18
             return 'Correct'
          return 'Incorrect'
20
21
22
       def make\_correction\_di(x):
23
          if 'dicoccum' in x['Sample Type']:
24
             if x['res'] != 4:
25
                return 'Correct'
             else:
                return 'Incorrect'
28
          if x['res'] != 3:
29
             return 'Correct'
30
          return 'Incorrect'
31
32
       if dom:
33
          return (b,d)
       b['Prediction'] = b.apply(make correction mono, axis=1)
35
       d['Prediction'] = d.apply(make_correction_di, axis=1)
36
       return (b,d)
37
```

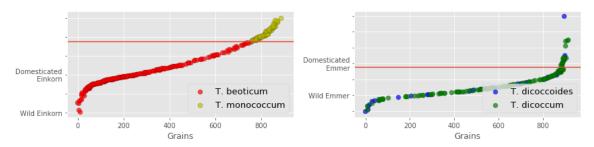
Make Test Plot 1

```
b, d = make_test_dfs()
fig, axes = plt.subplots(1,2)
   _ = sns.countplot(data = b, x='Sample Type', hue='Prediction', ax=axes[0])
   _ = sns.countplot(data = d, x='Sample Type', hue='Prediction', ax=axes[1])
   _ = axes[0].set_title('Einkorn')
   _ = axes[1].set_title('Emmer')
```

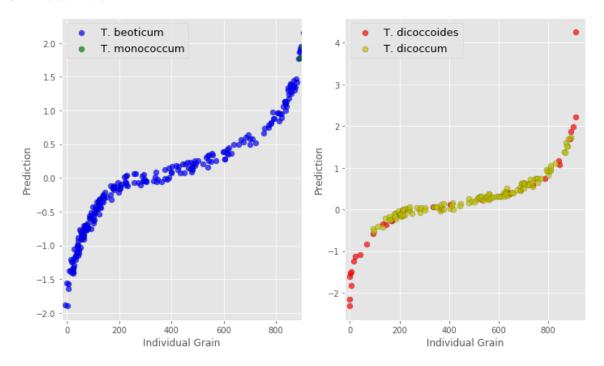
Make Test Plot 2

```
b, d = make test dfs(dom=True)
    fig, axes = plt.subplots(1,2, sharey=True)
2
     c = iter(['r', 'g', 'b', 'y'])
       = \text{sns.regplot}(\text{data} = \text{b[b['ypred']} == 1], x='\text{dom'}, y='\text{ypred'},
6
                  ax=axes[0], fit_reg=False, scatter_kws={"s": 50, "linewidth":0.2, "edgecolors":'k', 'alpha':0.7},
                  label=u,x jitter=10, y jitter=0.1, color= next(c))
        = \text{sns.regplot}(\text{data} = \text{b[b['ypred']} == 2], x='dom', y='ypred',
10
                  ax=axes[0], fit reg=False, scatter kws={"s": 50, "linewidth":0.2, "edgecolors":'k', 'alpha':0.7},
11
                  label=u,x jitter=10, y jitter=0.1, color= next(c))
12
13
        = sns.regplot(data = d[d['ypred'] == 3], x='dom', y='ypred',
14
                  ax=axes[1], fit_reg=False, scatter_kws={"s": 50, "linewidth":0.2, "edgecolors":'k', 'alpha':0.7},
15
                  label=u,x jitter=10, y jitter=0.1, color= next(c))
16
17
        = \text{sns.regplot}(\text{data} = \text{d}[\text{d}['\text{ypred}'] == 4], x = '\text{dom}', y = '\text{ypred}',
18
                  ax=axes[1], fit reg=False, scatter kws={"s": 50, "linewidth": 0.2, "edgecolors":'k', 'alpha':0.7},
19
                  label{label} = u, x\_jitter = 10, \ y\_jitter = 0.1, \ color = next(c))
20
        = axes[0].set title('Einkorn')
21
          axes[1].set_title('Emmer')
22
```

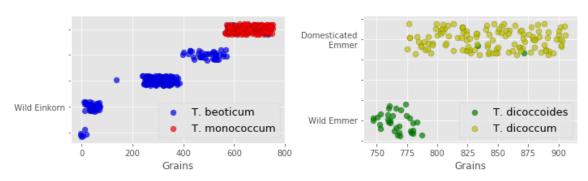
2.2.2 Model Original



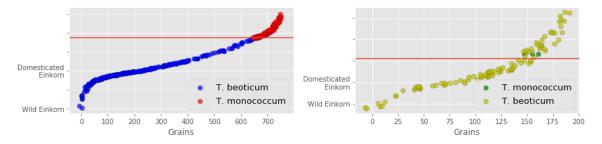
2.2.3 Model Bad



2.2.4 Model Flat

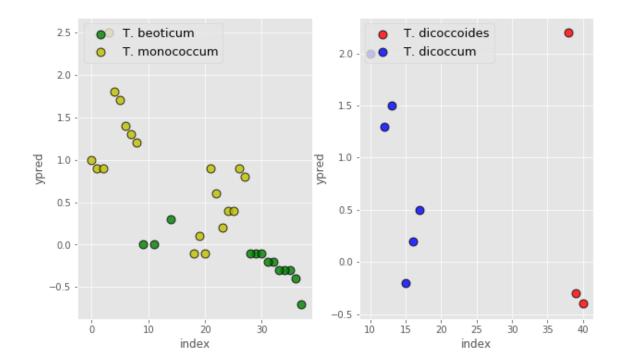


2.2.5 Model With Test Data

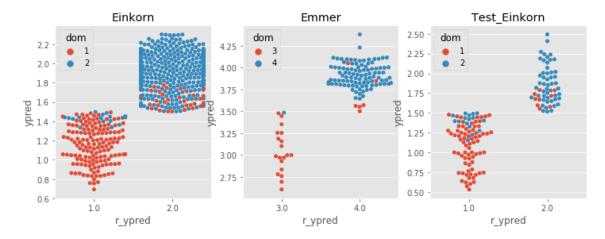


2.3 Predict apart

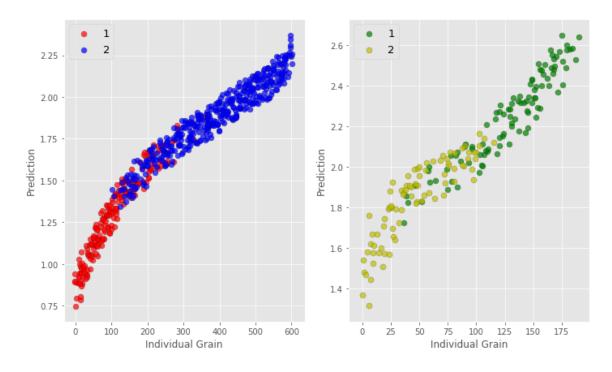
<matplotlib.legend.Legend at 0x7f0929f7bc50>



3 Test OLS



4 Bayesian Modelling



5 NN

