socioeconomic models

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```
[1]: # analytics
     import pandas as pd
     import numpy as np
     import scipy.stats as stats
     import statsmodels.formula.api as smf
     #spatial
     import osmnx as ox
     import geopandas as gpd
     import contextily as cx
     # plotting
     import seaborn as sns
     import matplotlib.pyplot as plt
     from matplotlib.colors import LinearSegmentedColormap
     #settings
     import warnings
     # set dataframe outputs to three digits
     pd.set_option("display.precision", 3)
     #suppress warnings
     warnings.filterwarnings('ignore')
[2]: # import data
     path = '/Users/philip/Documents/ESE/ESE_thesis/flood_experience/data/export/
      ⇔clean_n.csv'
     df_n = pd.read_csv(path)
     df_n.columns
[2]: Index(['id', 'state', 'zipcode', 'geographic_division', 'census_region',
            'county', 'experience', 'supplies', 'insured', 'involved',
            'learned_routes', 'made_plan', 'made_safer', 'planned_neighbors',
            'practiced_drills', 'documents', 'rainy_day', 'alerts',
            'family_communication', 'none', 'dont_know', 'age', 'sex', 'education',
            'race', 'homeownership', 'income', 'rentmortgage', 'rurality',
            'hazard_weight', 'geometry', 'zip_count'],
           dtype='object')
```

We will go through these outcome variables one by one structural adaptation: -made_safer

 $non-structural\ adaptation: -insurance - learned_routes - supplies - involved - made_plan - practiced_drills - alerts - family_communication$

```
[4]: def r_square(model):
    # McKelvay-Zavoina
    xb = model.predict(linear=True) #fitted latent value
    var_xb = np.var(xb,ddof=1) # variance of xb
    r2_mz = var_xb / (var_xb + 1) # McKelvay-Zavoina R_2
    # McFadden
    r2_mf = model.prsquared
    return r2_mz
```

```
[5]: def probit(functions, data):
         results list = []
         for func in functions:
             model = smf.probit(formula=func, data=data).fit(disp=0)
             df model = pd.DataFrame({
                 'effect': model.params,
                 'p': model.pvalues,
                 'pseudoR_2': r_square(model),
                 'LLPr': model.llr pvalue,
                 'BIC': model.bic
             })
             df_marginal = model.get_margeff().summary_frame()
             df_model = pd.concat([df_model, df_marginal], axis =1)
             df_model.index = pd.MultiIndex.from_product([[func], df_model.index],_
      →names=['function', 'beta'])
             results list.append(df model)
         results = pd.concat(results_list)
```

return results

```
[6]: #duplicate but with logit
     def logit(functions, data):
         results_list = []
         for func in functions:
             model = smf.logit(formula=func, data=data).fit(disp=0)
             marg_effects = model.get_margeff().summary_frame()
             df_model = pd.DataFrame({
                 'effect': model.params,
                 'p': model.pvalues,
                 'marginal_effect': marg_effects['dy/dx'],
                 'pseudoR_2': model.prsquared,
                 'LLPr': model.llr_pvalue,
                 'BIC': model.bic
             })
             df_model.index = pd.MultiIndex.from_product([[func], df_model.index],_
      ⇔names=['function', 'beta'])
             results_list.append(df_model)
         results = pd.concat(results_list)
         return results
```

```
[8]: results = probit(functions=functions, data=df_n)
results = results.round(3) # set to three decimal places
results.to_excel('results/probit_socioeconomic.xlsx')
```