

Supplementary Material Two

December 13, 2022

1 Overview

This is a python notebook to present all the model diagnostics of bayseian models used in Examining the Longitudinal Trajectory of Biopsychosocial Difficulties in Anorexia Nervosa Within a Bayesian Framework paper.

For more information on how the data was pre-processed and how the models where built please see https://github.com/WMDA/BB_data/

1.1 Content

Section 1: Overview of notebook

Section 2: All the linear and mixed regression models, libraries and data maniputaion needed for this notbook to run.

Section 3: A heatmap plot

Section 4: Kernel Density Estimation (KDE) and trace plots for the linear regression models

Section 5: KDE and trace plots for the mixed effects models

Section 6: Autocorrelation plots

2 Data manipulation and model fitting

This section can be skipped if model diagnostics is all that is wanted

```
[1]: from functions.data_functions import load_data, save_pickle, load_pickle
import seaborn as sns
sns.set_style('darkgrid')
import matplotlib.pyplot as plt
import pandas as pd
import warnings
warnings.filterwarnings(action='ignore')
import bambi as bmb
import arviz as az
```

2.1 Read in and preprocess data

```
[3]: pca_df = load_data('BEACON', 'pca_df')
comp = ['comp_1', 'comp_2', 'comp_3']
```

Demean the data

```
[4]: for component in comp:
    pca_df[f'{component}_t1_demean'] = pca_df[f'{component}_t1'] -_
    →pca_df[f'{component}_t1'].mean()

pca_df['group_ind'] = pca_df['group'].apply(lambda group: 0 if group == 'HC'_
    →else 1)
```

2.2 Linear regressions to test for differences at time point one

This is defining and fitting the linear regression models

```
[ ]: t1_models = {}

for component in comp:
    ttest = bmb.Model(f'{component}_t1 ~ 0 + group', data=pca_df.dropna())
    print(ttest)
    ttest_results = ttest.fit()
    t1_models[component] = ttest_results
```

2.3 Defining and fitting the mixed effects models.

For more detail on priors and deviations please see supplementary material one that is included with the paper

```
[7]: models = {
    'null':{},
    'alternative':{}
}

pca_df = pca_df.rename(columns={'G_Number':'participants'})
for model_component in comp:
    null = bmb.Model(f'{model_component}_t2 ~ 0 + (0 +_
    →{model_component}_t1_demean|participants)', data=pca_df, noncentered=False)
    alternative = bmb.Model(f'{model_component}_t2 ~ 0 +_
    →{model_component}_t1_demean + group + (0 +_
    →{model_component}_t1_demean|group)', data=pca_df, noncentered=False)

    models['null'][f'{model_component}'] = null
    models['alternative'][f'{model_component}'] = alternative
```

```
[ ]: for alternative_model in models['alternative'].keys():
    print(models['alternative'][alternative_model], '\n')

[ ]: for null_models in models['null'].keys():
    print(models['null'][null_models], '\n')

[10]: draw_numb = 2000
tune_numb = 3500
target_accept_numb = 0.95

[ ]: fitted_models ={
    'null':{},
    'alternative':{}
}

for alternative_model in models['alternative'].keys():
    alternative_fitted = models['alternative'][alternative_model].
    ↪fit(draws=draw_numb, tune=tune_numb, target_accept=target_accept_numb)
    print('\n')
    fitted_models['alternative'][alternative_model] = alternative_fitted

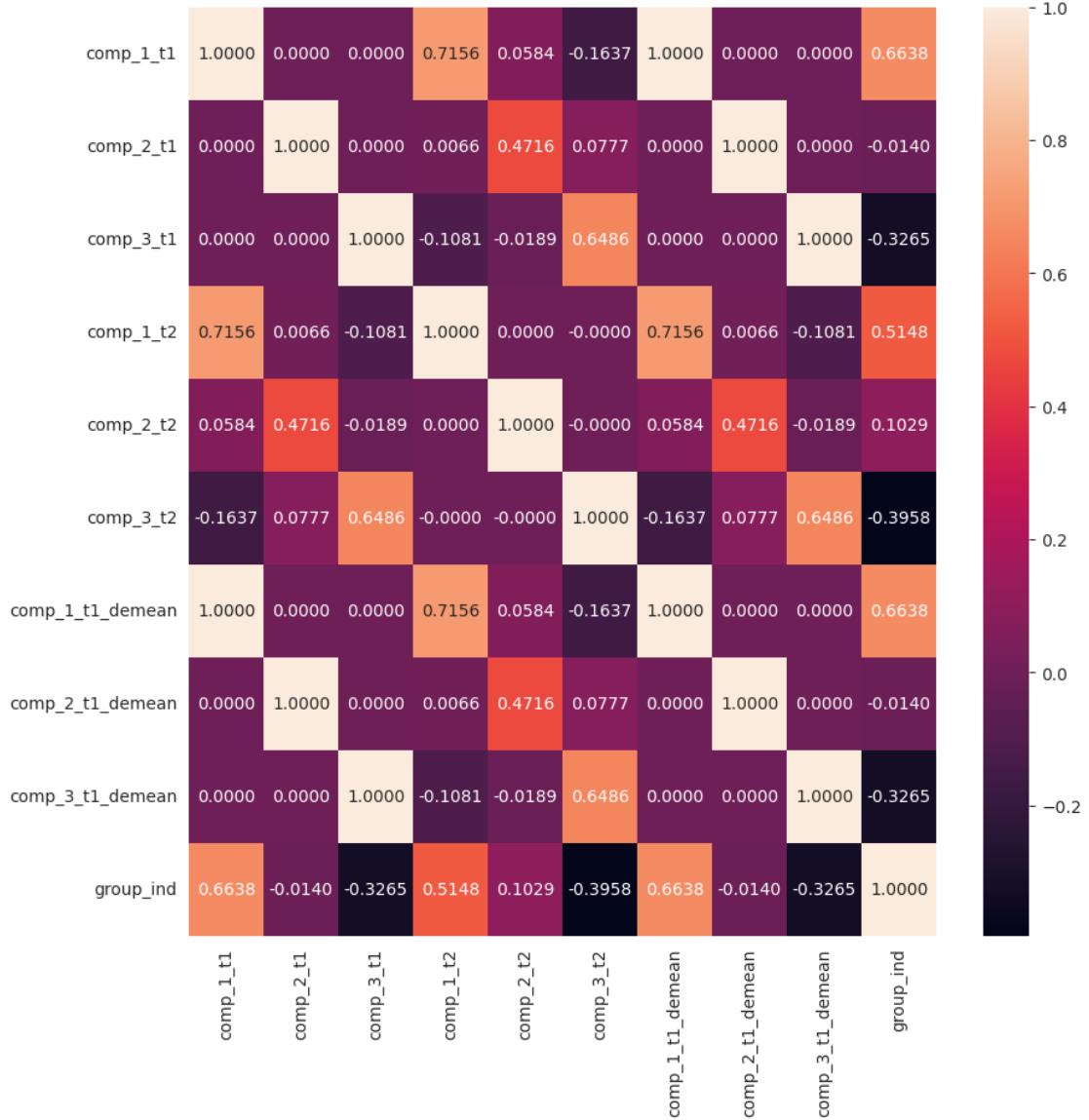
for null_model in models['null'].keys():
    null_fitted = models['null'][null_model].fit(draws=draw_numb, ↪
    ↪tune=tune_numb, target_accept=target_accept_numb)
    print('\n')
    fitted_models['null'][null_model] = null_fitted
```

3 Heatmap

This is a heatmap of all the variables in the used in the models to check for multicollinearity

```
[ ]: plt.figure(figsize=(10, 10))
sns.heatmap(pca_df[['comp_1_t1', 'comp_2_t1', 'comp_3_t1', 'comp_1_t2', ↪
    ↪'comp_2_t2', 'comp_3_t2', 'comp_1_t1_demean', 'comp_2_t1_demean', ↪
    ↪'comp_3_t1_demean', 'group_ind']].corr(), annot=True, fmt=".4f")
```

<AxesSubplot: >



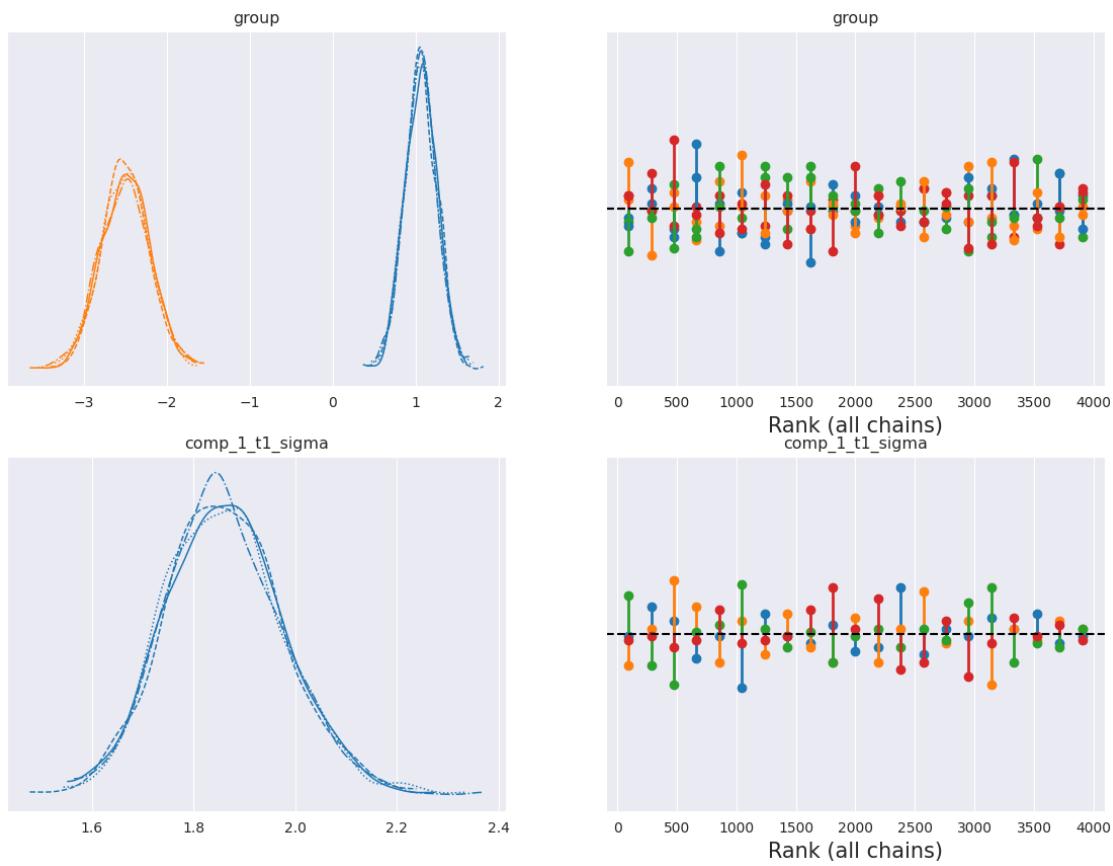
4 The KDE and the trace plots for linear regressions at time point one

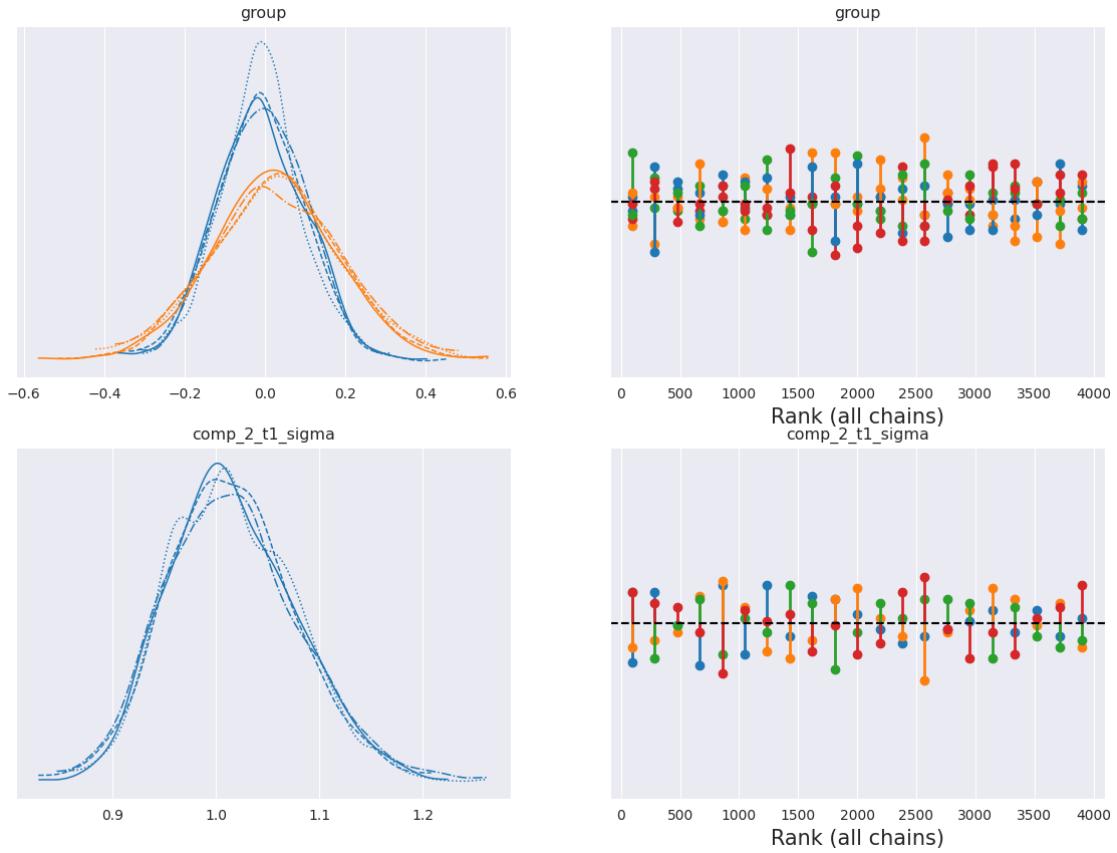
```
[22]: for model in t1_models.keys():
    print(model, '\n')
    az.plot_trace(t1_models[model], kind="rank_vlines", figsize=(14,10))
```

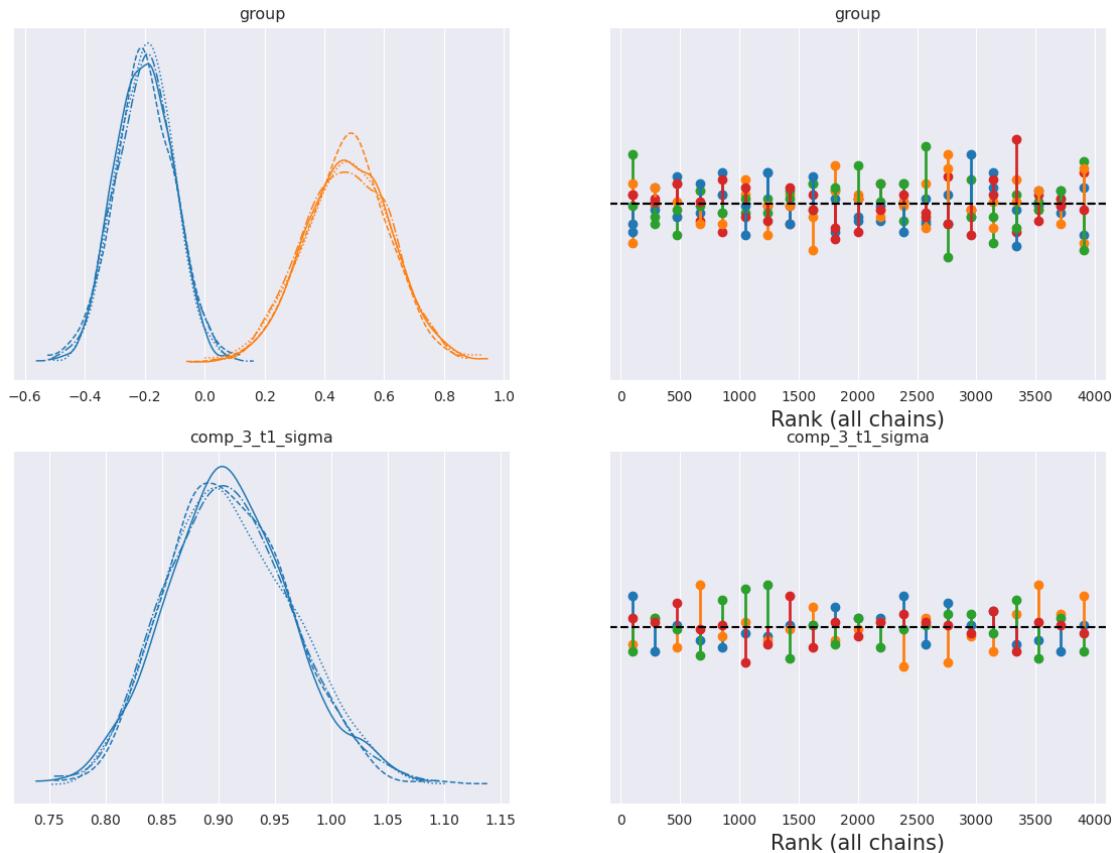
comp_1

comp_2

comp_3



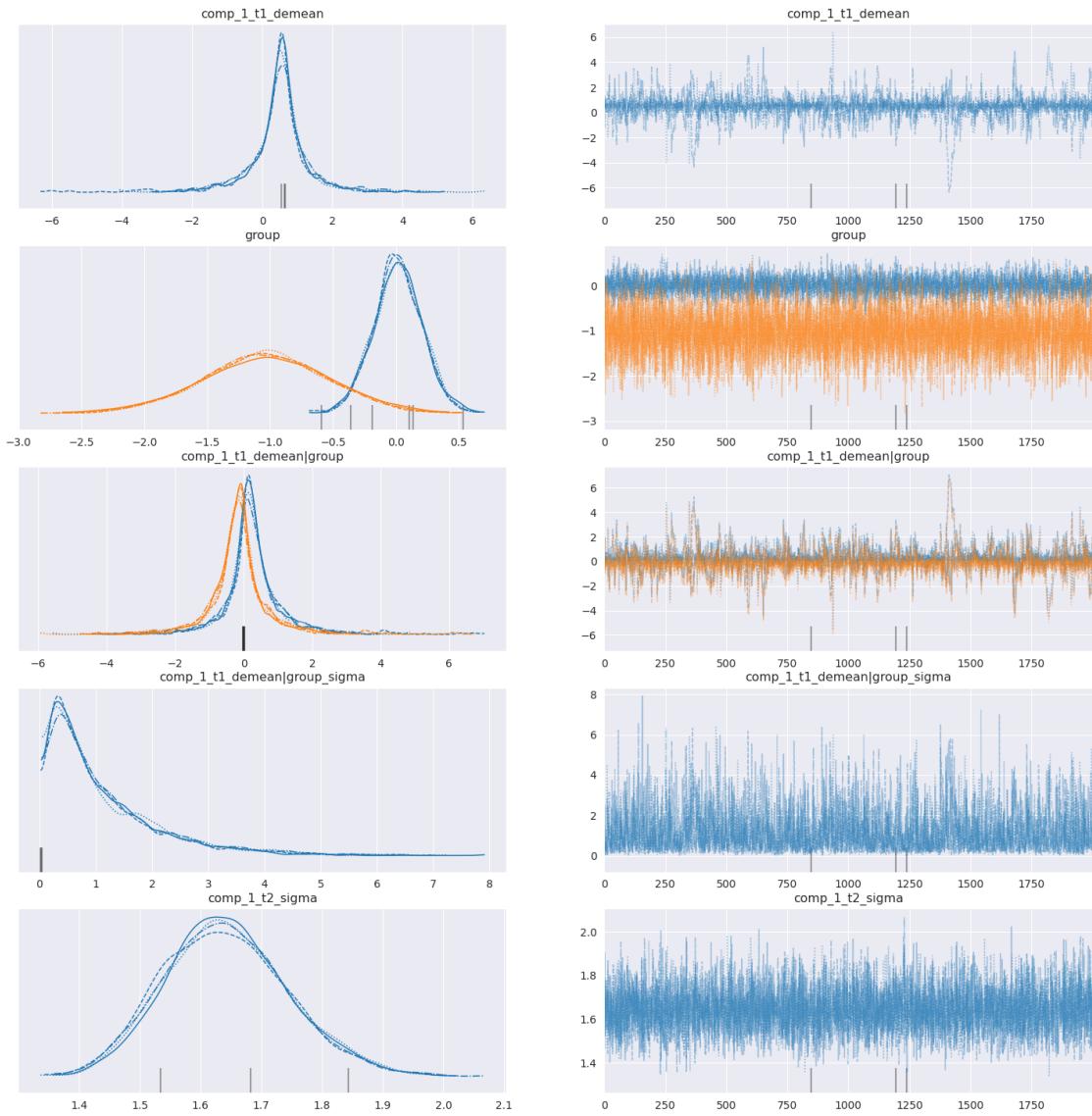


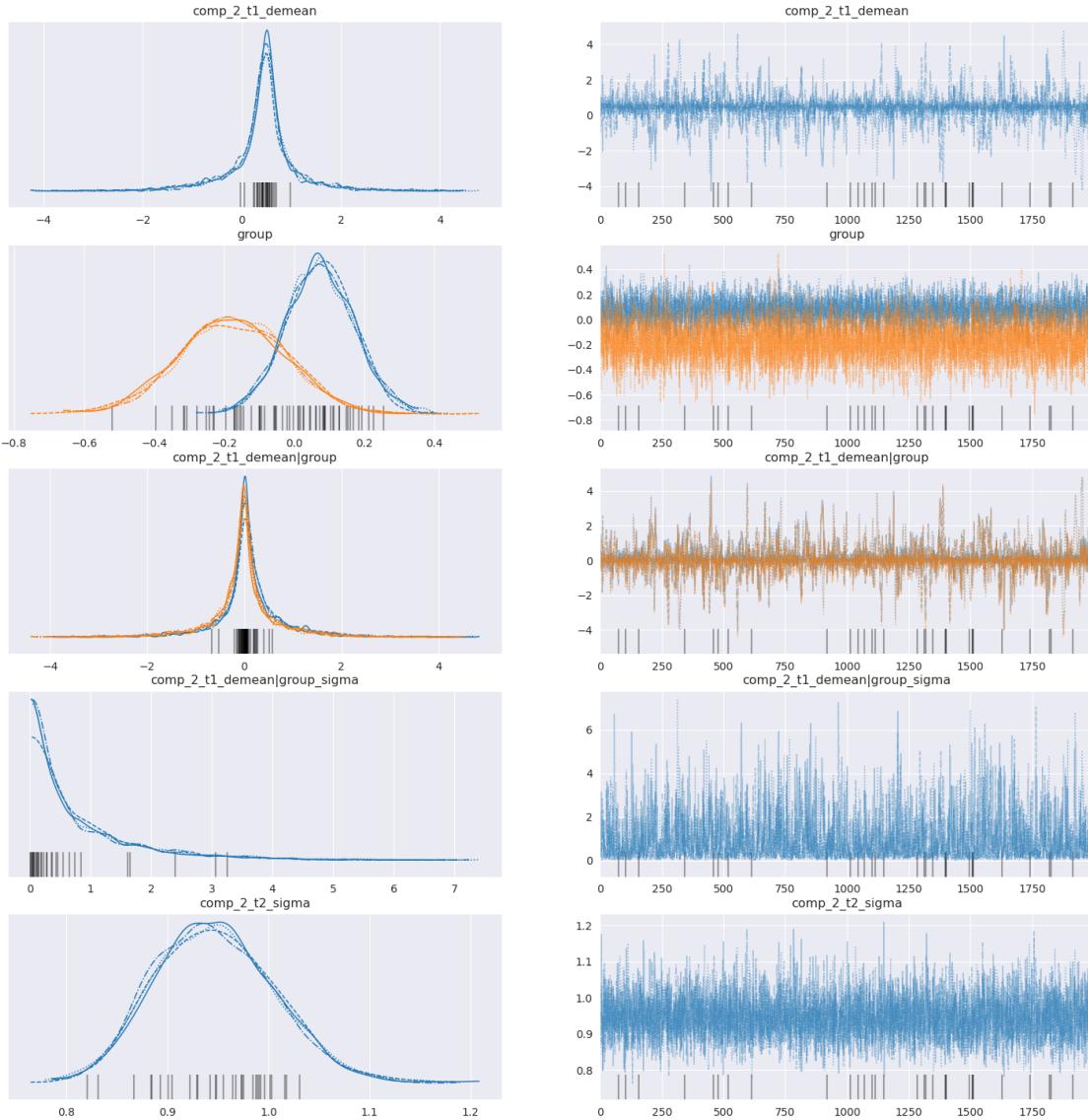


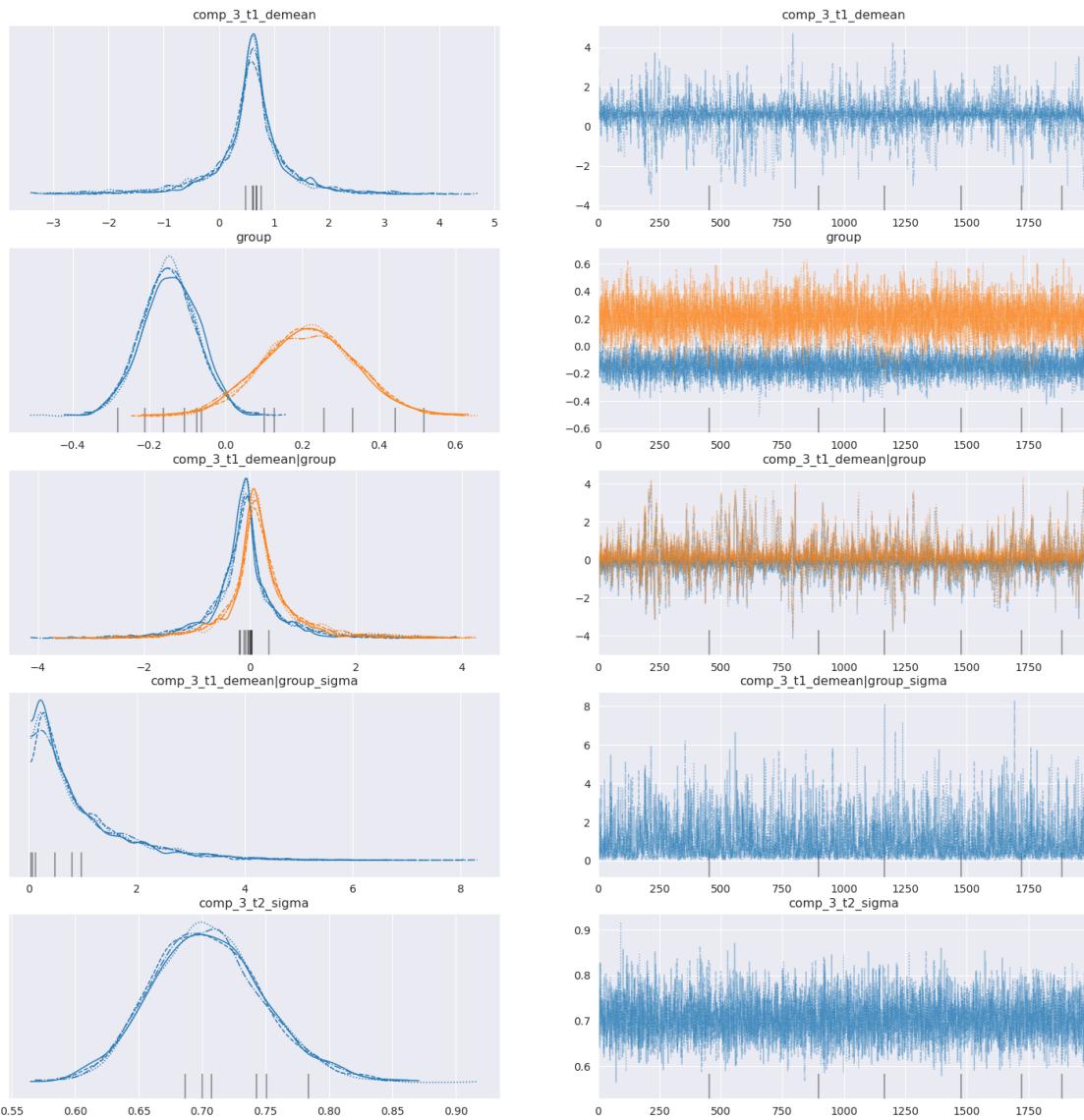
5 KDE and trace plots for mixed effects models

5.1 Plotting the KDE and the trace plots for the alternative model.

```
[14]: for model in comp:
    az.plot_trace(fitted_models['alternative'][model], figsize=(18,18))
```

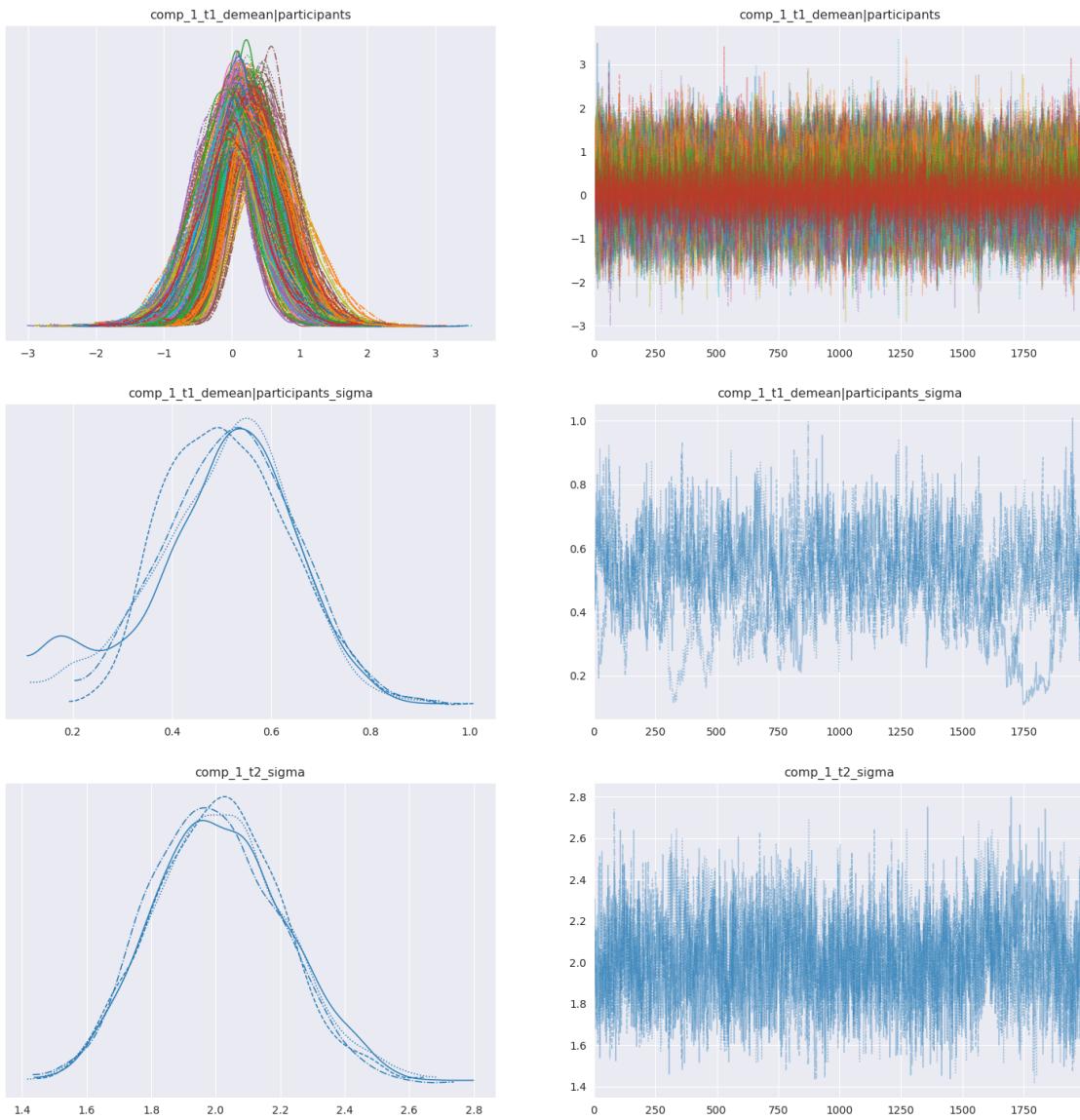


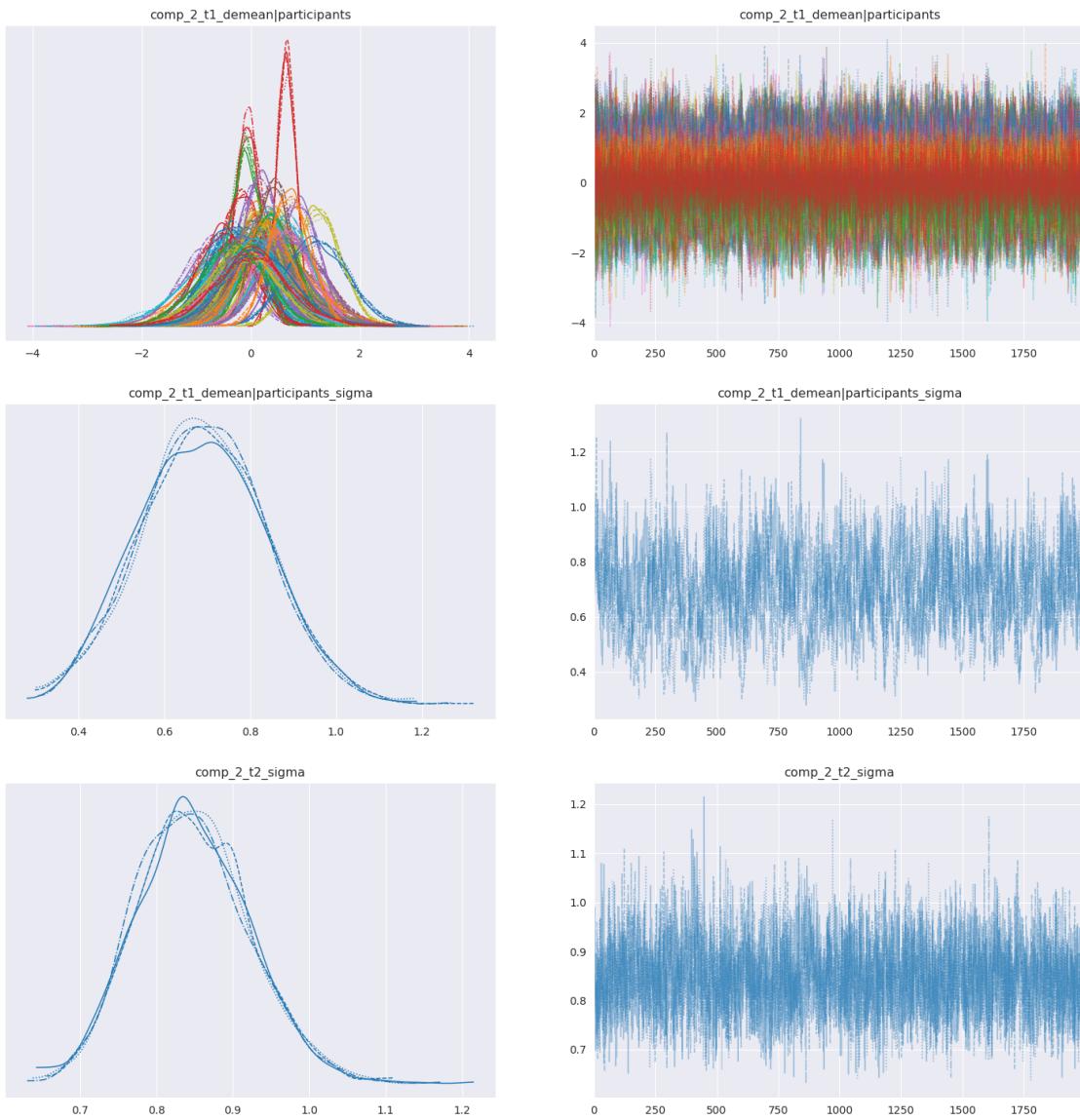


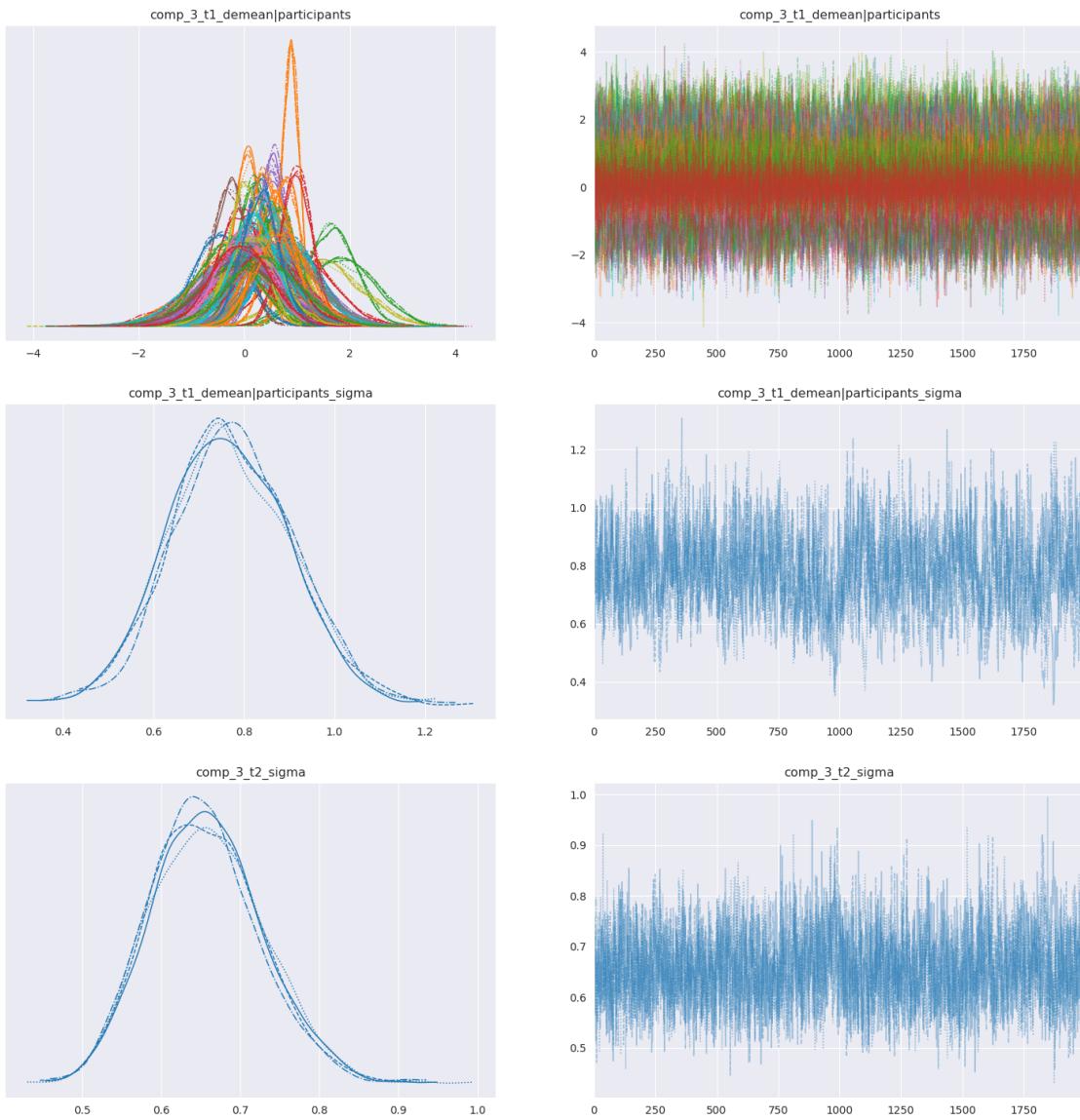


5.2 Plotting the KDE and trace plots for the null hypothesis models

```
[15]: for model in comp:
    az.plot_trace(fitted_models['null'][model], figsize=(18,18))
```



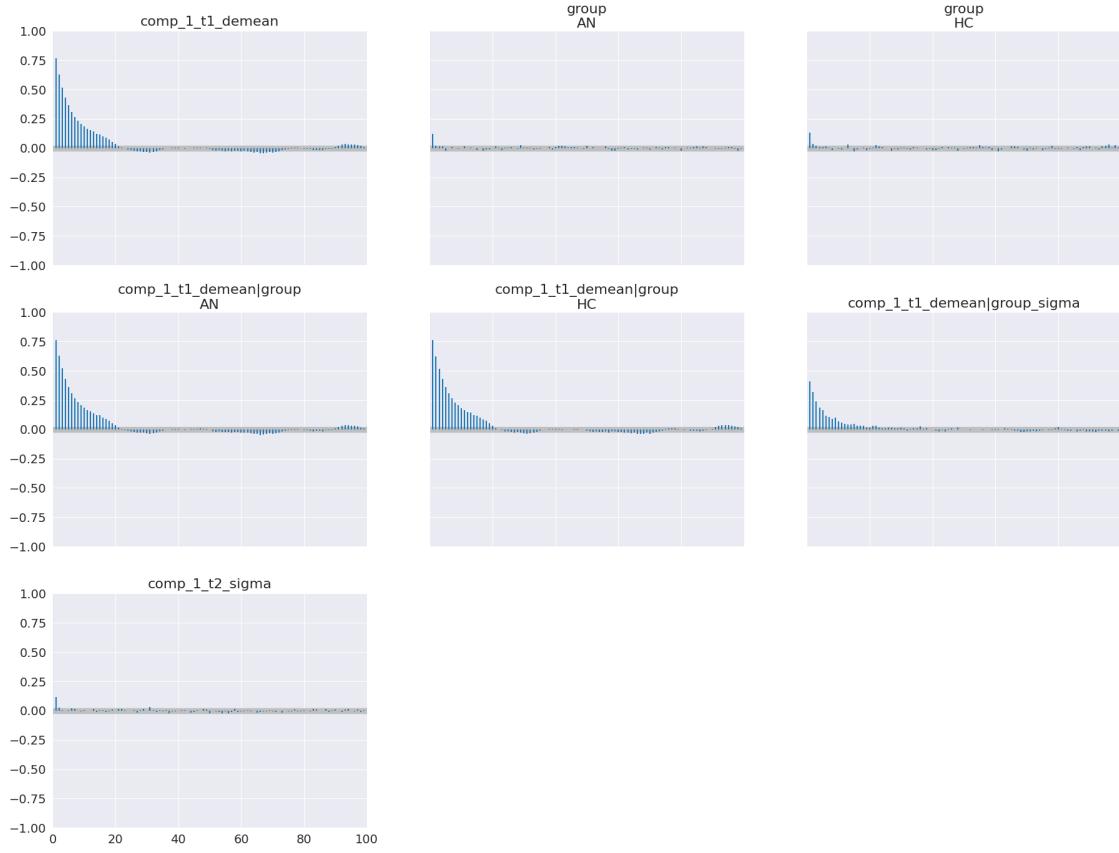


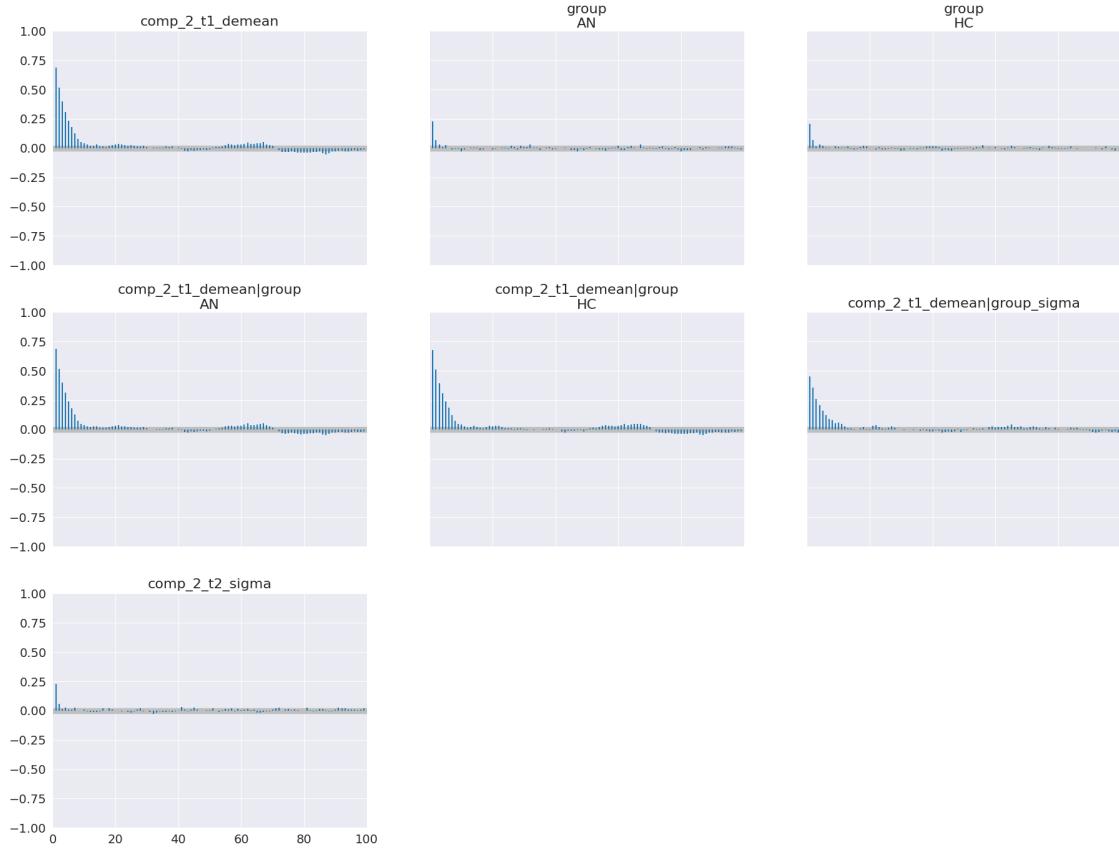


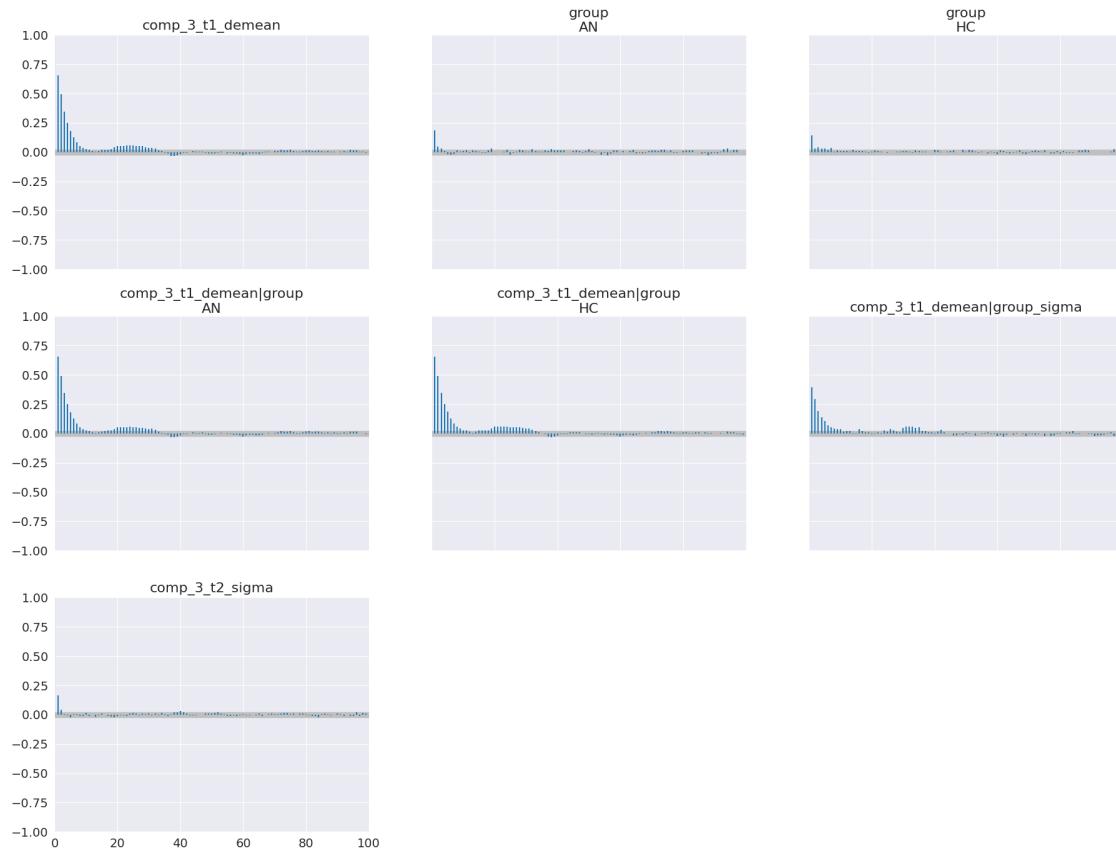
6 Autocorrelation plots

6.1 Autocorrelation plots for the alternative model

```
[12]: for model in comp:
    az.plot_autocorr(fitted_models['alternative'][model], combined=True)
```







6.2 Autocorrelation plots for null model

```
[13]: for model in comp:
    az.plot_autocorr(fitted_models['null'][model], combined=True)
```

