```
In [96]:
          import pandas as pd
          import seaborn as sns
          import matplotlib.pyplot as plt
          import os
          from sys import platform
          import numpy as np
 In [97]: df1 = pd.read excel("Data/Web Teaser Bewertung 01.xlsx")#, sheet name=None)
 In [98]: df2 = pd.read excel("Data/Web Teaser Bewertung 02.xlsx")#, sheet name=None)
 In [99]: df3 = pd.read excel("Data/Web Teaser Bewertung 03.xlsx")#, sheet name=None)
In [100]: dfs = [df1, df2, df3]
In [101]: | target_names = ["accent colour", "background colour", "font colour", "font c
          ontrast", "space", "roundness"]
In [102]: def split input target(df):
              header = df.iloc[0]
              df = df.rename(columns=df.iloc[0]).iloc[1:201]
              if platform == "linux" or platform == "linux2":
                  df = df.drop(columns="teaser_name")
              target_names = ["accent colour", "background colour", "font colour", "fo
          nt contrast", "space", "roundness"]
              targets = df[target names]
              inputs = df.drop(columns=target_names)
              inputs = inputs.astype(int)
              return inputs, targets
In [103]: inputs = []
          targets = []
          for df in dfs:
              input_df, target_df = split_input_target(df)
              inputs.append(input_df)
              targets.append(target_df)
In [104]: inputs1, targets1 = split input target(df1)
          inputs2, targets2 = split input target(df2)
```

Check data integrity of target

```
In [108]: #targets2[mask]
```

Show deviation of responses:

```
In [109]: # Inpect data:
inputs[2]
```

Out[109]:

	Innovative - Traditional	Elegant - Lassig	Emotional - Sachlich	Jung - Erfahren	Perfektionistisch - Spontan	zurückhalten - selbstsicher	Weiblich - Männlich
1	-6	5	-10	-5	7	2	-8
2	1	2	2	4	2	5	6
3	-8	4	-6	-6	8	7	-6
4	-3	4	4	-2	3	-7	3
5	-5	3	-7	-7	6	-4	-4
				•••			
196	6	-4	6	8	-6	-6	6
197	-5	6	-5	-2	8	8	6
198	3	6	7	6	3	-6	8
199	3	-3	6	6	-6	0	6
200	-3	-3	-3	3	6	0	8

200 rows × 7 columns

```
In [110]: # Calculate mean
    mean_input = inputs[0].copy()
    for inp in inputs[1:]:
        mean_input += inp.copy()
    mean_input /= len(inputs)
```

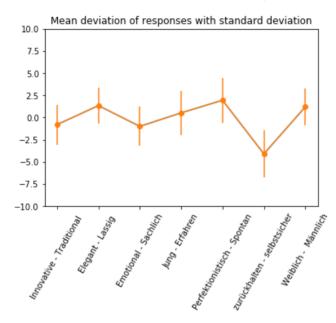
```
In [111]: mean_input.mean()
```

```
Out[111]: Innovative - Traditional -0.803333
Elegant - Lassig 1.336667
Emotional - Sachlich -1.005000
Jung - Erfahren 0.520000
Perfektionistisch - Spontan 1.938333
zurückhalten - selbstsicher 4.108333
Weiblich - Männlich 1.186667
dtype: float64
```

```
In [112]: # Calculate std:
    std_input = (inputs[0].copy() - mean_input) ** 2
    for inp in inputs[1:]:
        std_input += (inp.copy() - mean_input) ** 2
    std_input /= len(inputs)
    std_input = std_input ** 0.5
```

```
In [113]: std_input.mean()
Out[113]: Innovative - Traditional
                                          2,259430
                                          1.998557
          Elegant - Lassig
          Emotional - Sachlich
                                          2.192584
          Jung - Erfahren
                                          2.502932
          Perfektionistisch - Spontan
                                         2.502957
          zurückhalten - selbstsicher
                                         2.707655
          Weiblich - Männlich
                                          2.069644
          dtype: float64
In [114]:
          plt.plot(mean input.mean())
          plt.errorbar(range(len(mean_input.mean())), mean_input.mean(),yerr=std_inpu
          t.mean(), fmt='-o')
          plt.ylim(-10, 10)
          plt.xticks(rotation=60)
          plt.title("Mean deviation of responses with standard deviation")
```

Out[114]: Text(0.5, 1.0, 'Mean deviation of responses with standard deviation')



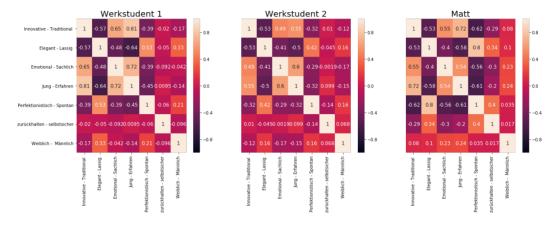
Correlation between input columns

```
In [115]: corr1 = inputs[0].corr()
    corr2 = inputs[1].corr()
    corr3 = inputs[2].corr()
    corr_total = inputs[0].append(inputs[1]).append(inputs[2]).corr()
```

```
In [116]: fig = plt.figure(figsize = (20, 20)) # width x height
    ax1 = fig.add_subplot(331) # row, column, position
    ax2 = fig.add_subplot(332)
    ax3 = fig.add_subplot(333)

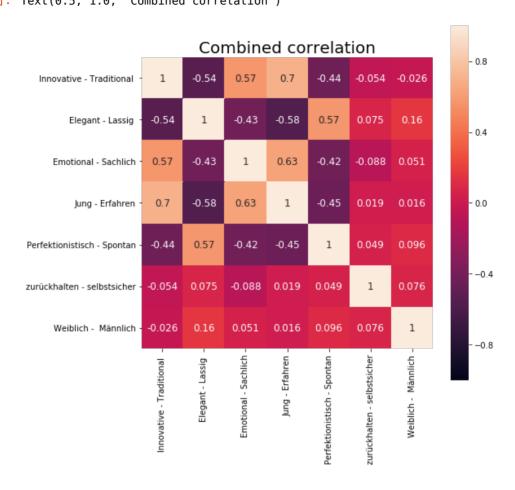
sns.heatmap(corr1, ax=ax1, vmin=-1, vmax=1, annot=True, annot_kws={'fontsize': 12}).set_title('Werkstudent 1', fontsize =20)
    sns.heatmap(corr2, ax=ax2, vmin=-1, vmax=1, yticklabels=False, annot=True, annot_kws={'fontsize': 12}).set_title('Werkstudent 2', fontsize =20)
    sns.heatmap(corr3, ax=ax3, vmin=-1, vmax=1, yticklabels=False, annot=True, annot_kws={'fontsize': 12}).set_title('Matt', fontsize =20)
```

Out[116]: Text(0.5, 1.0, 'Matt')



```
In [117]: fig = plt.figure(figsize = (8, 8)) # width x height
ax1 = fig.add_subplot(111) # row, column, position

sns.heatmap(corr_total, ax=ax1, vmin=-1, vmax=1, square=True, annot=True, an
not_kws={'fontsize': 12}).set_title('Combined correlation', fontsize =20)
Out[117]: Text(0.5, 1.0, 'Combined correlation')
```



Sklearn tests:

```
In [118]:
           targets[0]
Out[118]:
                accent colour background colour font colour font contrast
                                                                  space roundness
                                              #1e2722
              1
                    #d24041
                                       #ffffff
                                                                   tiaht
                                                                               1
                                                          medium
                    #7e0068
                                              #121212
              2
                                    #ededed
                                                          medium
                                                                    tight
                                                                               1
              3
                    #131313
                                       #ffffff
                                              #555555
                                                             low
                                                                 medium
              1
                    #e05e00
                                     #fef9f4
                                              #121212
                                                             low
                                                                    tight
                    #c70200
                                     #f7f6f7
              5
                                              #121212
                                                          medium
                                                                medium
                     #ff502d
            196
                                     #efecea
                                              #000000
                                                          medium medium
            197
                    #9299a0
                                       #ffffff
                                              #2e3439
                                                             low
                                                                medium
            198
                    #546276
                                     #02234f
                                               #ffc3bc
                                                          medium medium
                                                                             NaN
                    #005h55
            199
                                     #fdd6d0
                                              #005c55
                                                          medium medium
                                                                               1
            200
                                    #000000
                       NaN
                                                 #ffffff
                                                          medium
                                                                               4
                                                                    tiaht
           200 rows × 6 columns
In [119]: # Check NANs
           pd.concat(targets).isnull().mean()
Out[119]: accent colour
                                  0.265
           background colour
                                  0.005
           font colour
                                  0.000
           font contrast
                                  0.000
                                  0.000
           space
           roundness
                                  0.040
           dtype: float64
In [120]: | dataset = pd.concat((pd.concat(inputs), pd.concat(targets)), axis=1)
In [121]: dataset = dataset.dropna(subset=["roundness"])
In [122]: inputs_np = dataset.drop(columns=target_names).to_numpy()
           inputs_np.shape
Out[122]: (576, 7)
In [123]:
           # preprocess:
           from sklearn.preprocessing import StandardScaler
           inputs np = StandardScaler().fit transform(inputs np)
In [124]: pd.options.mode.chained_assignment = None # default='warn''
In [125]: # Extract target values that we want to predict:
           used_targets = ["roundness"]
           \#used\_targets = ["space", "font contrast", "roundness"]
           targets = dataset[target names]
           targets["space"] = targets["space"].astype("category").cat.codes
           targets["font contrast"] = targets["font contrast"].astype("category").cat.c
           targets["roundness"] = targets["roundness"].astype("category").cat.codes
           targets_np = targets[used_targets].to_numpy()
In [126]: from sklearn.model selection import train test split
```

```
In [127]:
          import sklearn
          from sklearn.svm import SVC
          from sklearn.neural_network import MLPClassifier
          from sklearn.ensemble import RandomForestClassifier
          from sklearn.multioutput import MultiOutputClassifier
In [128]: print(inputs_np.shape, targets_np.shape)
          (576, 7) (576, 1)
In [129]: def apply_classifier(inputs, targets, classifier):
              X train, X test, Y train, Y test = train test split(inputs np, targets n
              classifier.fit(X train, Y train)
              Y pred = classifier.predict(X test)
              #print(sklearn.metrics.classification_report(Y_test, Y_pred))
              accuracy = classifier.score(X_test, Y_test)
              print("Accuracy: ", accuracy)
              return classifier
In [130]: #classifier = SVC()
          classifier = MultiOutputClassifier(SVC())
          classifier = apply classifier(inputs np, targets np, classifier)
          Accuracy: 0.84027777777778
          /home/anton/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:193: Fu
          tureWarning: The default value of gamma will change from 'auto' to 'scale' in
          version 0.22 to account better for unscaled features. Set gamma explicitly to
          'auto' or 'scale' to avoid this warning.
            "avoid this warning.", FutureWarning)
In [131]: | classifier = MultiOutputClassifier(MLPClassifier(solver='adam', max iter=100
          0, batch_size=32))
          classifier = apply_classifier(inputs_np, targets_np, classifier)
          Accuracy: 0.868055555555556
In [132]: classifier = MultiOutputClassifier(RandomForestClassifier())
          classifier = apply_classifier(inputs_np, targets_np, classifier)
          Accuracy: 0.8958333333333334
          /home/anton/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:
          245: FutureWarning: The default value of n_estimators will change from 10 in
          version 0.20 to 100 in 0.22.
            "10 in version 0.20 to 100 in 0.22.", FutureWarning)
 In [ ]:
```