```
In [1]:
```

```
import pandas as pd
import numpy as np
```

### load files from data folder

#### In [2]:

```
log = pd.read_csv(r"C:\Users\s.goodroe\Dropbox\PhD\Teaching\exampleproject\data\log.cs
v")
response = pd.read_csv(r"C:\Users\s.goodroe\Dropbox\PhD\Teaching\exampleproject\data\re
sponse.csv")
correct = pd.read_csv(r"C:\Users\s.goodroe\Dropbox\PhD\Teaching\exampleproject\data\cor
rect.csv")
```

# create copy of response data and add metrics from the log file you want to use

```
In [3]:
```

```
data = response
```

#### In [4]:

```
data['female'] = log.gender
data['age'] = log.age
data['exp'] = log.experience
data['native'] = log.native_speaker
```

#### In [5]:

```
log
```

#### Out[5]:

	ppid	date	time	age	gender	native_speaker	experience
0	P_01	30/01/2020	10:20	18	m	у	24
1	P_02	30/01/2020	10:40	20	f	у	36
2	P_03	30/01/2020	11:00	21	f	у	20
3	P_04	30/01/2020	11:20	21	m	у	14
4	P_05	30/01/2020	11:40	29	m	у	10
5	P_06	30/01/2020	12:40	21	m	n	40
6	P_07	30/01/2020	13:40	20	f	у	19
7	P_08	30/01/2020	14:40	22	f	у	20
8	P_09	30/01/2020	15:40	24	m	n	26
9	P_10	31/01/2020	16:40	25	f	у	15

# replace non-numeric values with coded values (i.e. F = 1, Y = 1)

```
In [6]:
```

```
data = data.replace({'m':0,'f':1, 'y':1, 'n':0})
```

#### In [7]:

data #view dataset

#### Out[7]:

	ppid	RQ	ES	CG	TS	HP	RP	LS	СТ	SP	female	age	exp	native
0	P_01	289	105	112	315	119	219	190	26	218	0	18	24	1
1	P_02	33	250	143	358	298	137	215	81	10	1	20	36	1
2	P_03	266	299	6	20	146	233	182	81	348	1	21	20	1
3	P_04	26	105	302	86	165	160	280	78	103	0	21	14	1
4	P_05	94	52	67	173	296	300	277	65	320	0	29	10	1
5	P_06	151	143	115	274	157	197	233	287	248	0	21	40	0
6	P_07	241	17	330	252	79	169	214	225	180	1	20	19	1
7	P_08	336	177	60	178	59	74	199	222	103	1	22	20	1
8	P_09	351	159	156	171	168	217	112	5	253	0	24	26	0
9	P_10	104	165	143	201	336	319	264	0	6	1	25	15	1

# create output frame where you can log values such as stats you want

```
In [8]:
```

```
output = pd.DataFrame(columns = ['mean','sd'])
```

#### In [9]:

```
output['mean'] = data.mean()
output['sd'] = data.std()
```

```
In [10]:
```

```
output
```

#### Out[10]:

	mean	sd			
RQ	189.1	122.512993			
ES	147.2	84.688711			
CG	143.4	101.978429			
TS	202.8	102.057500			
HP	182.3	95.615027			
RP	202.5	73.221506			
LS	216.6	50.933944			
СТ	107.0	101.061038			
SP	178.9	120.297271			
female	0.5	0.527046			
age	22.1	3.142893			
exp	22.4	9.500877			
native	0.8	0.421637			

# how do you want to measure things such as accuracy / deviation from correct angle?

```
In [11]:
```

```
deviation = pd.DataFrame(columns = data.columns[0:10])
deviation['ppid'] = data.ppid
```

#### In [12]:

```
def get_dev(ppid,location):
    a = correct[location]
    resp = data[location].where(data.ppid == ppid).dropna()
    dev = resp.values - a
    return dev.values[0]
```

#### In [13]:

```
for i in range(len(data.ppid)):
    for location in correct.columns:
        deviation['ppid'][i] = data['ppid'][i]
        deviation[location][i] = get_dev(data.ppid[i],location)
```

#### In [14]:

```
deviation['av_dev'] = deviation.iloc[:,1:].mean(axis = 1)
deviation['sd_dev'] = deviation.iloc[:,1:].std(axis = 1)
```

#### In [15]:

deviation

#### Out[15]:

	ppid	RQ	ES	CG	TS	HP	RP	LS	СТ	SP	av_dev	sd_dev
0	P_01	230	-12	-196	-14	42	48	68	-177	20	1.000000	121.137204
1	P_02	-26	133	-165	29	221	-34	93	-122	-188	-6.555556	131.198841
2	P_03	207	182	-302	-309	69	62	60	-122	150	-0.333333	186.263255
3	P_04	-33	-12	-6	-243	88	-11	158	-125	-95	-31.000000	110.125181
4	P_05	35	-65	-241	-156	219	129	155	-138	122	6.666667	152.760524
5	P_06	92	26	-193	-55	80	26	111	84	50	24.555556	90.024825
6	P_07	182	-100	22	-77	2	-2	92	22	-18	13.666667	79.677126
7	P_08	277	60	-248	-151	-18	-97	77	19	-95	-19.555556	143.851861
8	P_09	292	42	-152	-158	91	46	-10	-198	55	0.888889	144.458717
9	P_10	45	48	-165	-128	259	148	142	-203	-192	-5.111111	161.563503

# add the mean deviation and sd for each location to your output

### In [16]:

```
output['dev_mean'] = deviation.iloc[:,1:].mean(axis = 0)
output['dev_sd'] = deviation.iloc[:,1:].std(axis = 0)
```

### In [17]:

output

### Out[17]:

	mean	sd	dev_mean	dev_sd
RQ	189.1	122.512993	130.1	122.512993
ES	147.2	84.688711	30.2	84.688711
CG	143.4	101.978429	-164.6	101.978429
TS	202.8	102.057500	-126.2	102.057500
HP	182.3	95.615027	105.3	95.615027
RP	202.5	73.221506	31.5	73.221506
LS	216.6	50.933944	94.6	50.933944
СТ	107.0	101.061038	-96.0	101.061038
SP	178.9	120.297271	-19.1	120.297271
female	0.5	0.527046	NaN	NaN
age	22.1	3.142893	NaN	NaN
exp	22.4	9.500877	NaN	NaN
native	0.8	0.421637	NaN	NaN

# look at potential trends by eye

## In [18]:

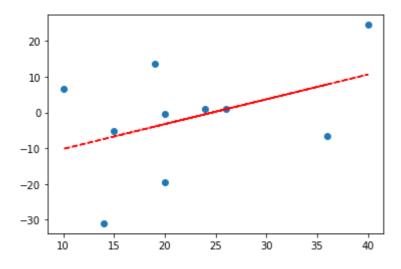
import matplotlib.pyplot as plt

#### In [19]:

```
x = data.exp
y = deviation.av_dev
fig = plt.scatter(x, y)
z = np.polyfit(x, y, 1)
p = np.poly1d(z)
plt.plot(x,p(x),"r--")
```

#### Out[19]:

[<matplotlib.lines.Line2D at 0x27016ebce48>]



## fit linear regression with n predictors

### In [20]:

```
import statsmodels.api as sm
```

```
In [21]:
```

```
x = data[['exp','age','female']]
y = deviation.av_dev

model = sm.OLS(y,x).fit()
predictions = model.predict(x)
x = sm.add_constant(x,has_constant = 'add')

model.summary()
```

C:\Users\s.goodroe\AppData\Local\Continuum\anaconda3\lib\site-packages\num py\core\fromnumeric.py:2389: FutureWarning: Method .ptp is deprecated and will be removed in a future version. Use numpy.ptp instead.

return ptp(axis=axis, out=out, \*\*kwargs)

C:\Users\s.goodroe\AppData\Local\Continuum\anaconda3\lib\site-packages\sci
py\stats\stats.py:1416: UserWarning: kurtosistest only valid for n>=20 ...
continuing anyway, n=10

"anyway, n=%i" % int(n))

#### Out[21]:

#### **OLS Regression Results**

R-squared (uncentered): 0.145 Dep. Variable: av\_dev Model: OLS Adj. R-squared (uncentered): -0.222 Method: Least Squares F-statistic: 0.3948 Date: Sun. 02 Feb 2020 Prob (F-statistic): 0.761 Time: 16:00:37 Log-Likelihood: -40.485 No. Observations: 10 AIC: 86.97 **Df Residuals:** 7 BIC: 87.88 Df Model: 3 nonrobust

Covariance Type: nonrobust

coef std err t P>|t| [0.0]

[0.025 0.975] 0.4741 0.487 0.973 0.363 -0.678 1.626 exp -0.4149 0.554 -0.749 0.478 -1.724 0.895 age female -5.0465 10.222 -0.494 0.637 -29.218 19.125

 Omnibus:
 0.935
 Durbin-Watson:
 2.325

 Prob(Omnibus):
 0.627
 Jarque-Bera (JB):
 0.582

 Skew:
 -0.530
 Prob(JB):
 0.747

 Kurtosis:
 2.477
 Cond. No.
 62.4

#### Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.