

Analizando a qualidade do meu sono

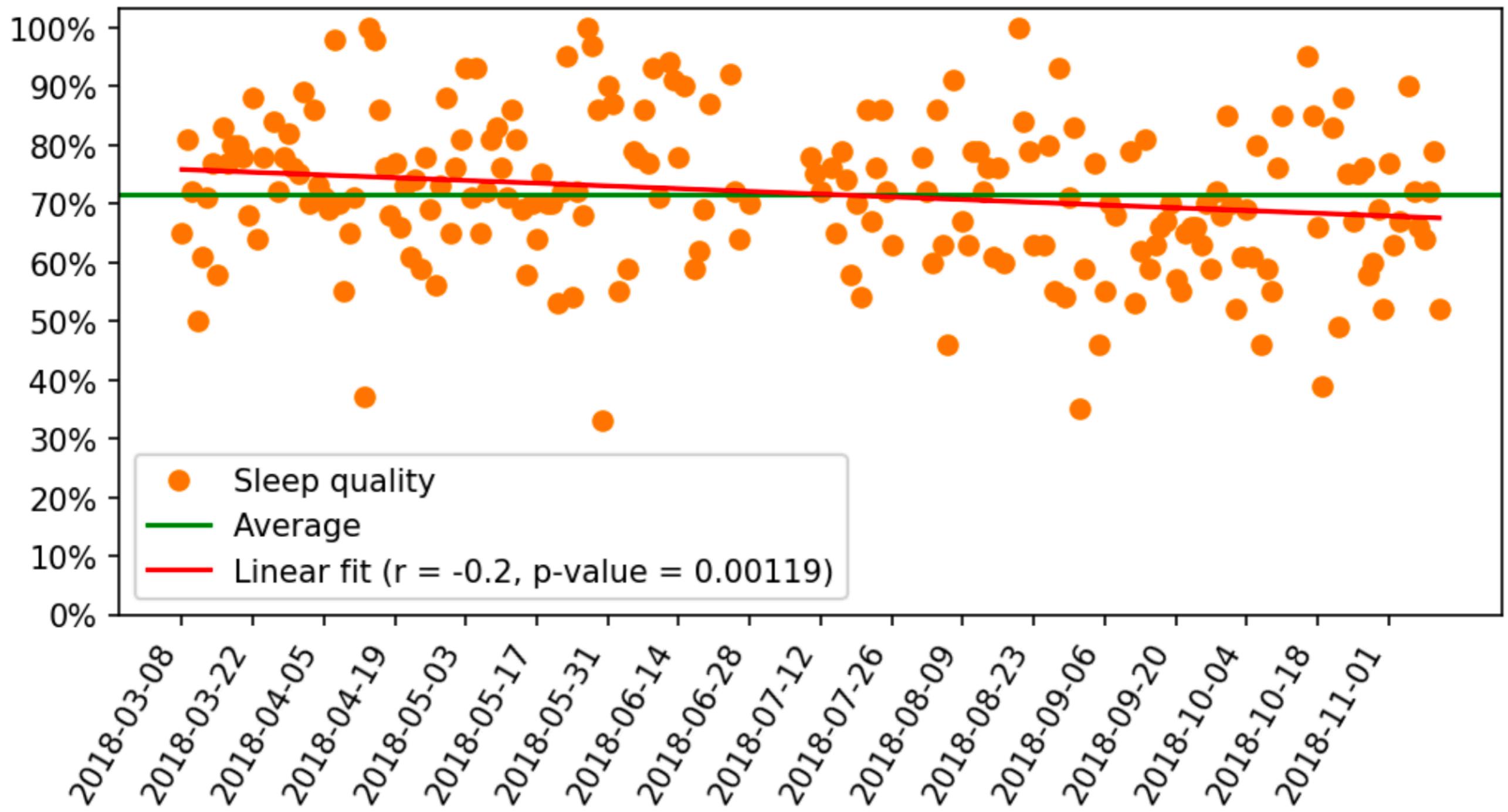
Reuben Nascimento Morais

Fontes de dados:

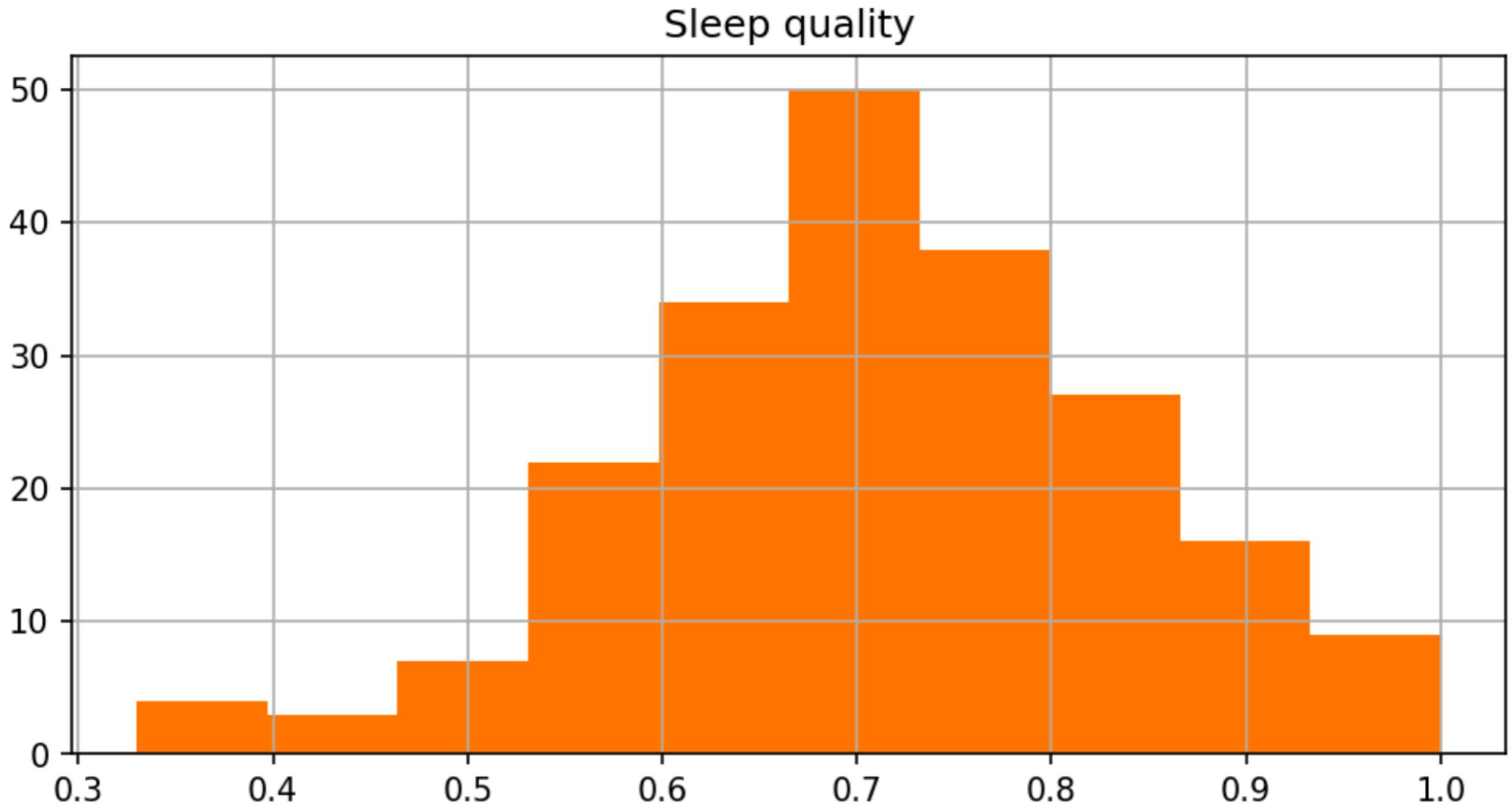
- [Sleep Cycle](#) (qualidade do sono, horas de sono)
- [Life Cycle](#) (tempo em casa/em aula/viajando)
- [Apple Health](#) (batimentos cardíacos por minuto, tempo de atividade física)
- [Fases da lua](#)

Dados de sono e localização coletados de 2018-03-07 a 2018-11-11 (232 dias).

Dados de batimentos cardíacos e esportes coletados de 2018-06-13 a 2018-11-11 (152 dias).

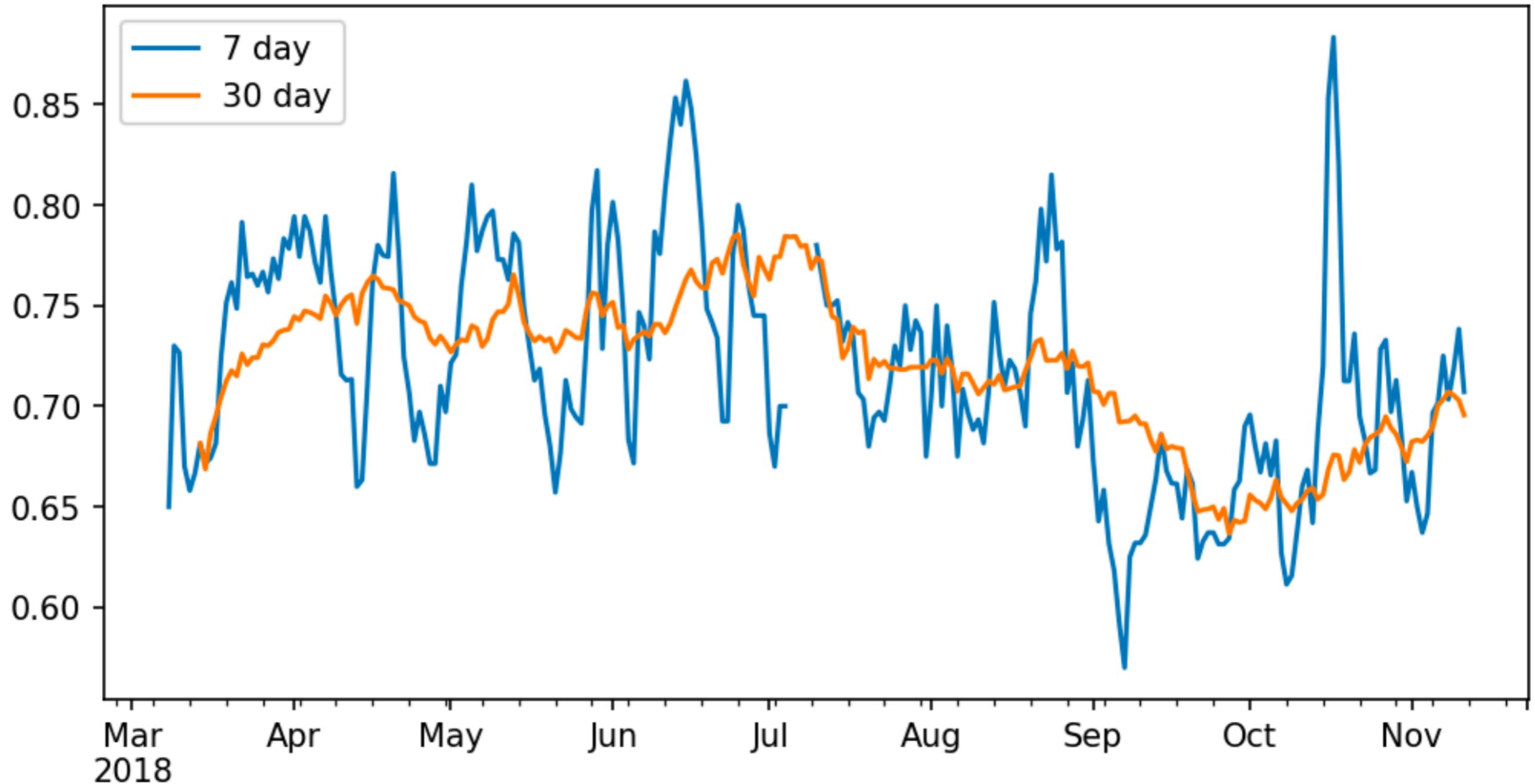


```
sleepquality.hist(color='C1')  
plt.title('Sleep quality')  
plt.show()
```



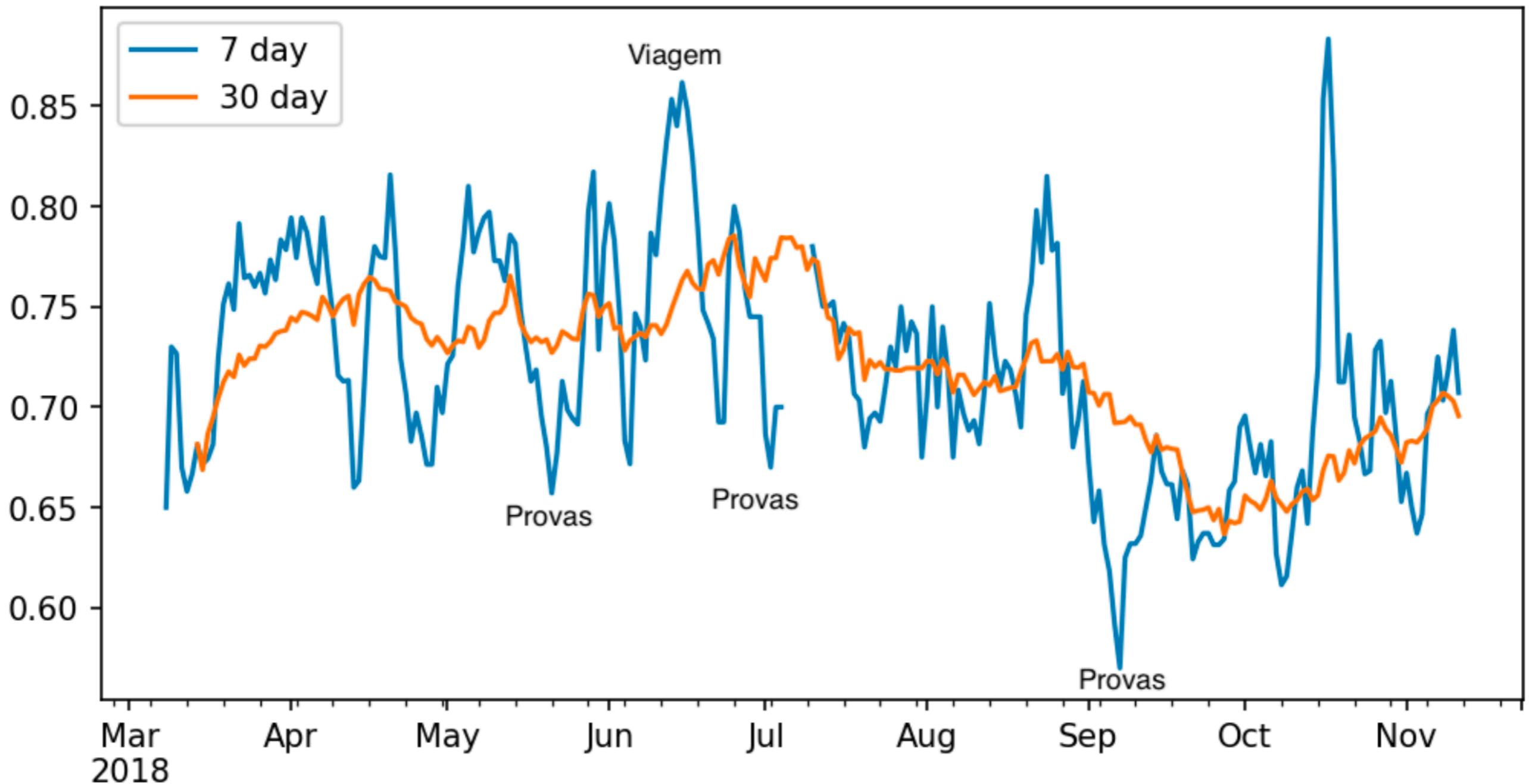
```
sleepquality.rolling('7d').mean().plot(label='7 day')
sleepquality.rolling('30d', min_periods=7).mean().plot(label='30 day')
plt.title('7 day and 30 day rolling mean sleep quality')
plt.legend()
plt.show()
```

7 day and 30 day rolling mean sleep quality



```
Image('annotated.png')
```

7 day and 30 day rolling mean sleep quality



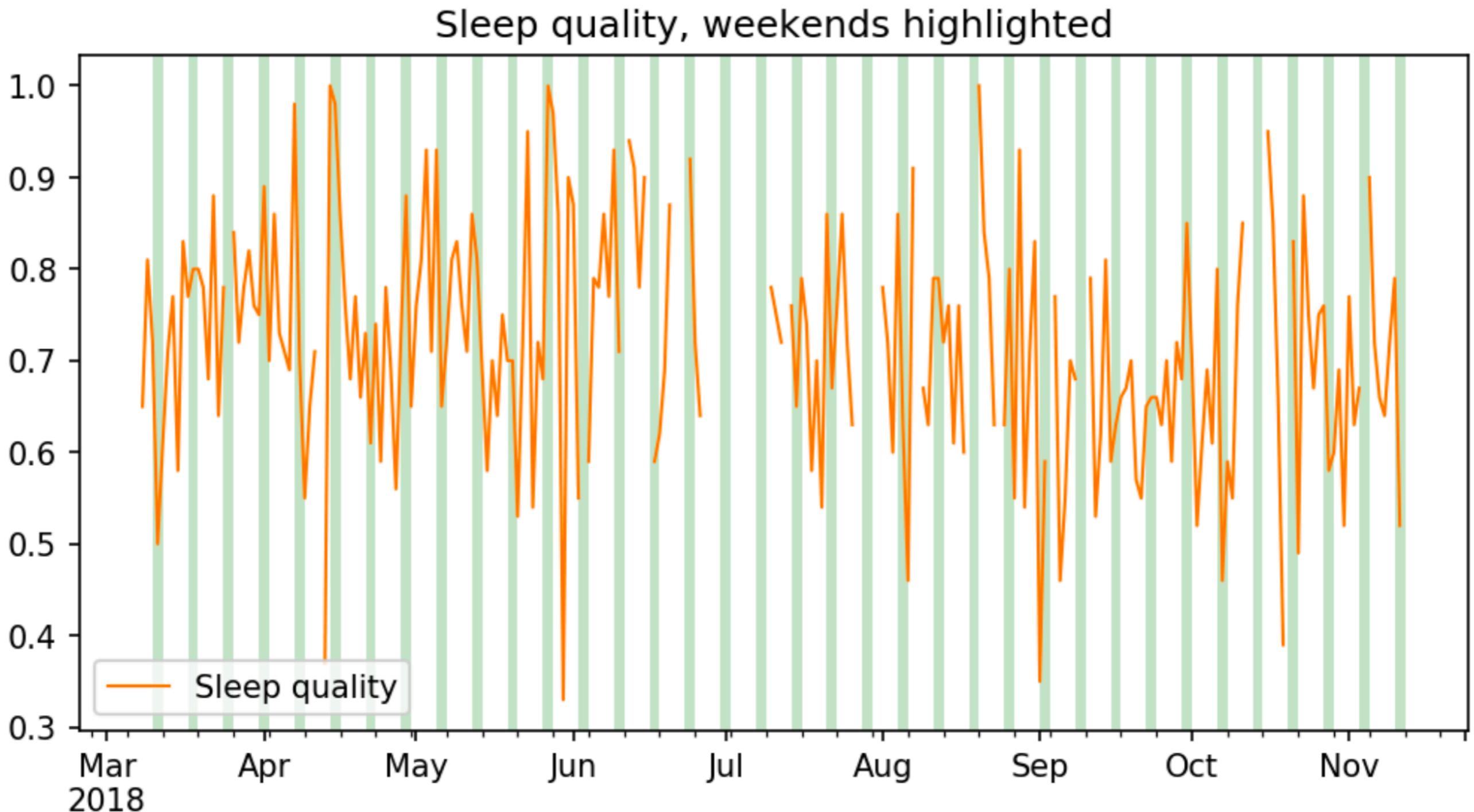
```

sleepquality.plot(color='C1', linewidth=1)

for i in sleepdata.index[sleepdata.index.weekday >= 5]:
    plt.gca().axvspan(i, i+pandas.to_timedelta('1d'), facecolor='green', edgecolor='none', alpha=.2)

plt.title('Sleep quality, weekends highlighted')
plt.legend(loc='lower left')
plt.show()

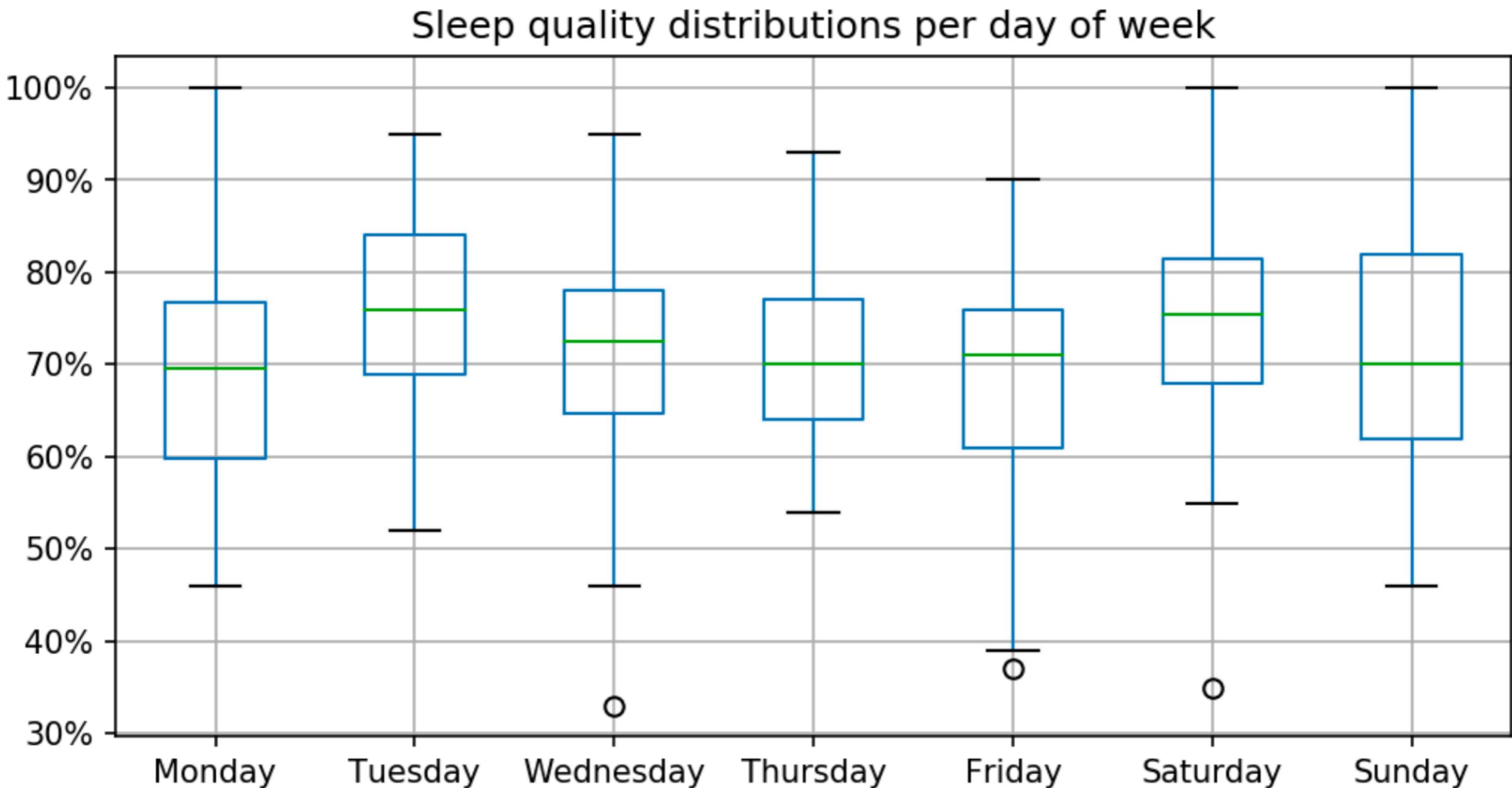
```



```
weekends = sleepquality[sleepquality.index.weekday >= 5]
weekdays = sleepquality[sleepquality.index.weekday < 5]
print(f'Global average: {quality_average}, n = {len(sleepdata)}')
print(f'Weekend average: {weekends.mean()}, n = {len(weekends)}')
print(f'Weekday average: {weekdays.mean()}, n = {len(weekdays)}')
print('P(quality > mean | weekend):', (weekends > quality_average).mean())
print('P(quality > mean | weekday):', (weekdays > quality_average).mean())
```

Global average: 0.7167142857142856, n = 249
Weekend average: 0.734181818181818, n = 72
Weekday average: 0.710516129032258, n = 177
P(quality > mean | weekend): 0.4027777777777778
P(quality > mean | weekday): 0.4350282485875706

```
weekdays_df = pandas.DataFrame(index=['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday'],  
                                data=[sleepquality[sleepquality.index.weekday == i].values for i in range(7)]).T  
weekdays_df.boxplot()  
plt.gca().yaxis.set_major_formatter(ticker.PercentFormatter(xmax=1))  
plt.title('Sleep quality distributions per day of week')  
plt.show()
```



Mean quality by moon phase:

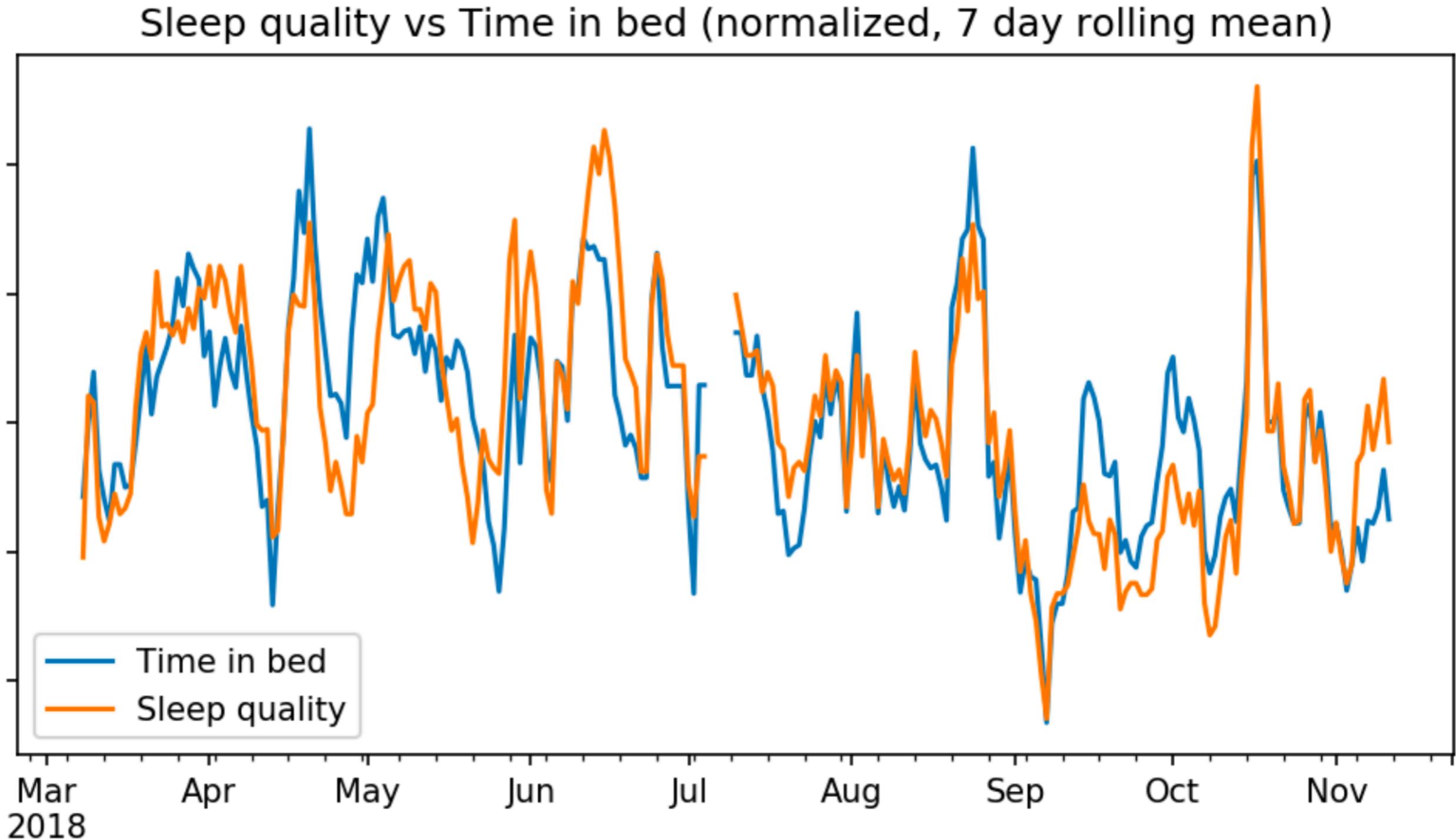
```
{'First Quarter': 0.7064583333333334,  
'Full Moon': 0.7230769230769233,  
'Last Quarter': 0.7018518518518518,  
'New Moon': 0.7339285714285715}
```

Global average: 0.7167142857142856

day>average vs. First Quarter: r = -0.0709, p-value = 0.265, n = 58
day>average vs. Full Moon: r = 0.0384, p-value = 0.547, n = 61
day>average vs. Last Quarter: r = -0.0494, p-value = 0.438, n = 65
day>average vs. New Moon: r = 0.0801, p-value = 0.208, n = 65

```
normalized_time.rolling('7d').mean().plot(color='C0')
normalized_quality.rolling('7d').mean().plot(color='C1')

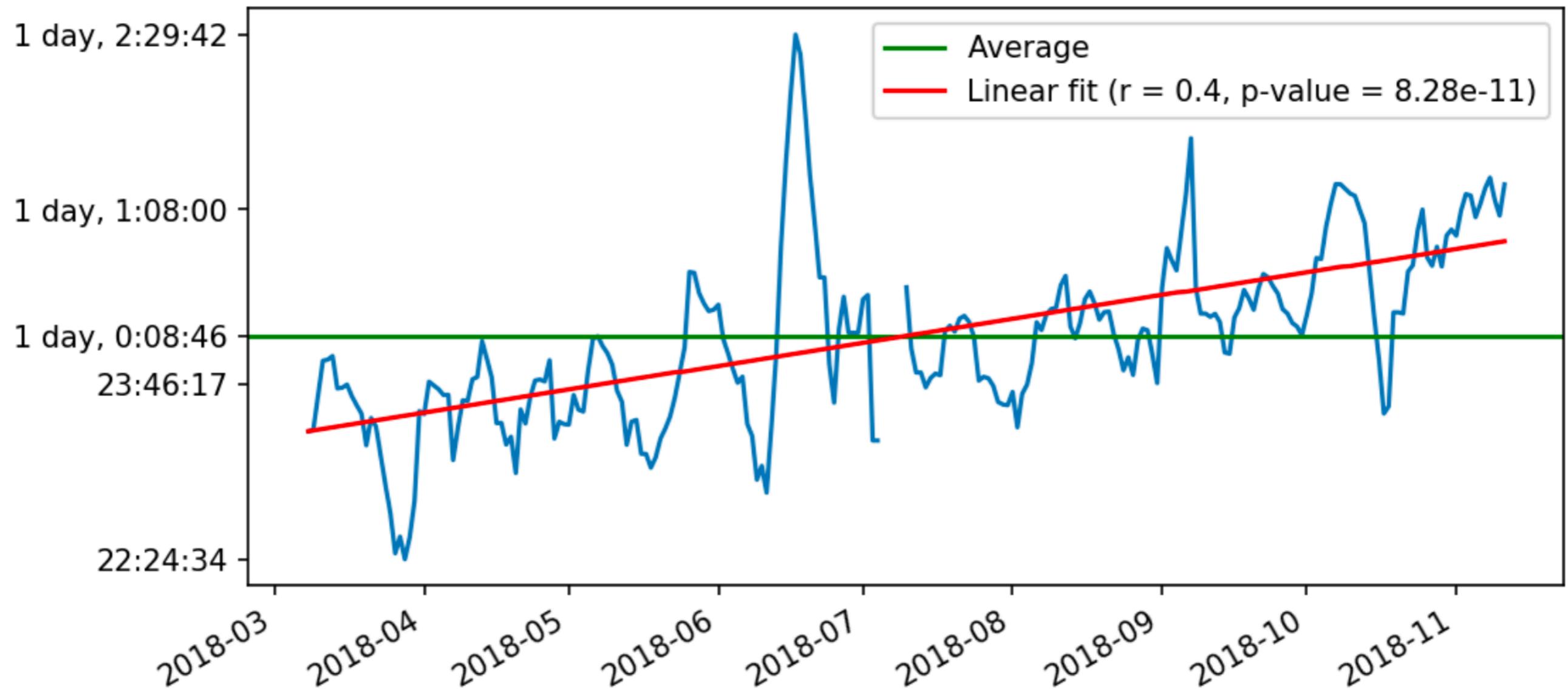
plt.title('Sleep quality vs Time in bed (normalized, 7 day rolling mean)')
plt.legend()
plt.show()
```



```
r, p = stats.pearsonr(normalized_time.fillna(0), normalized_quality.fillna(0))
print(f'r = {r}, p-value = {p}')
```

```
r = 0.8248180315592906, p-value = 4.164236371099999e-63
```

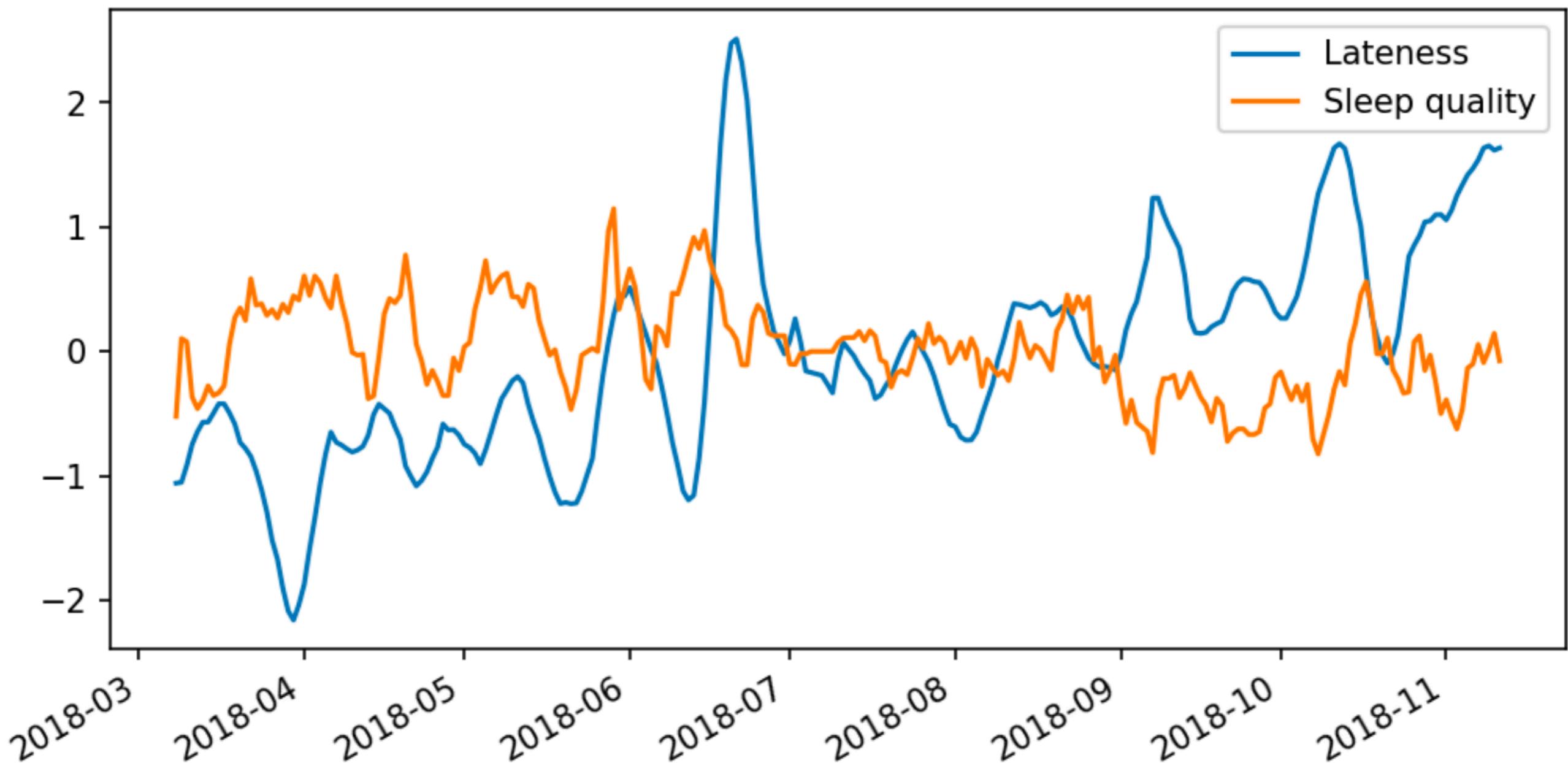
Lateness of start of sleep (7-day rolling mean)



```
aligned_lateness = align_and_compare(normalized_quality, normalized_lateness)
plt.title('Lateness vs sleep quality (7 day mean, normalized)')
plt.legend()
plt.show()
```

r = -0.19498935295208325, p-value = 0.0021244810286180703

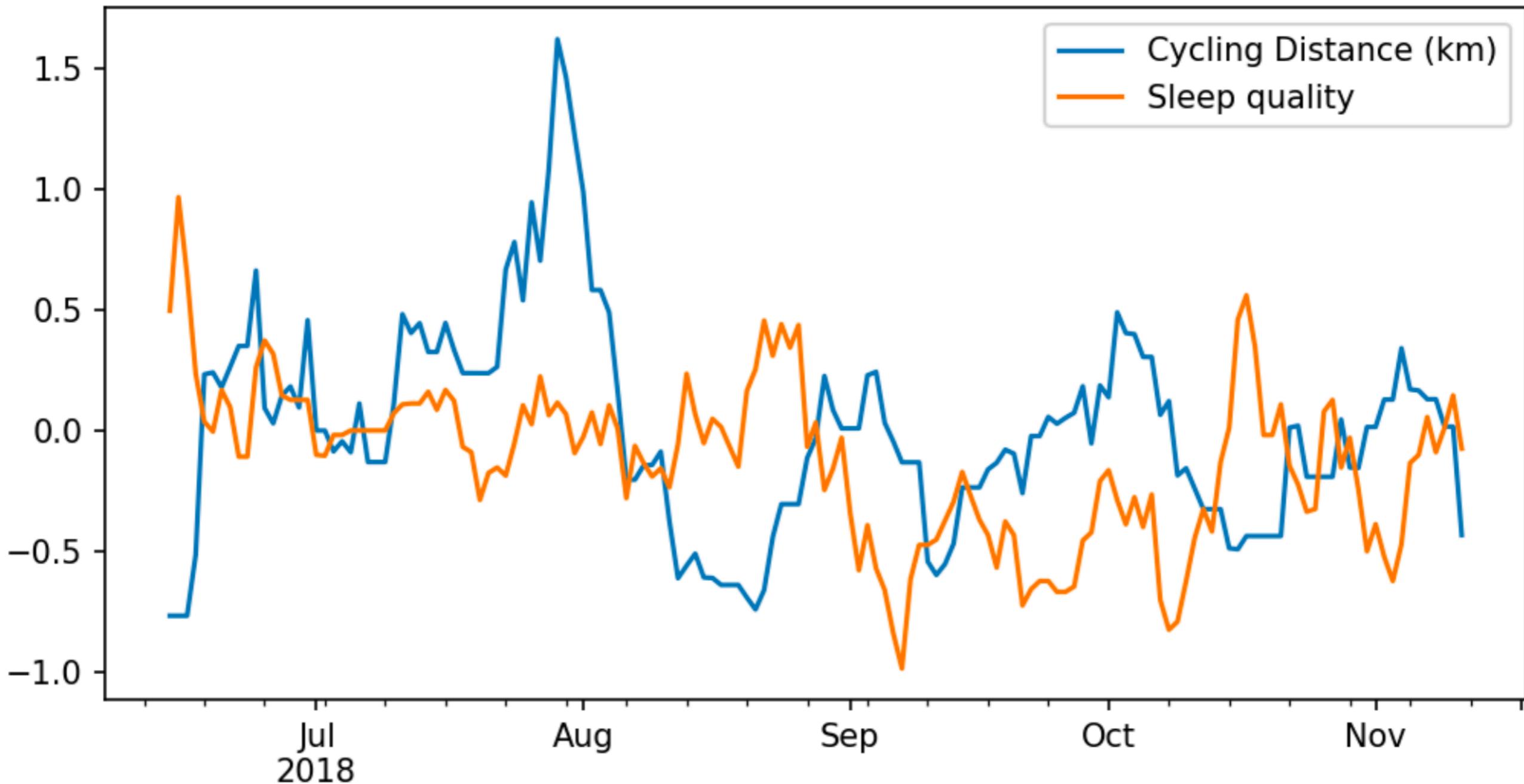
Lateness vs sleep quality (7 day mean, normalized)



```
aligned_cycling = align_and_compare(normalized_quality, normalized_cycling)
plt.title('Sleep quality vs cycling distance (7-day mean, normalized)')
plt.legend()
plt.show()
```

r = -0.062472522639179254, p-value = 0.4460357077804835

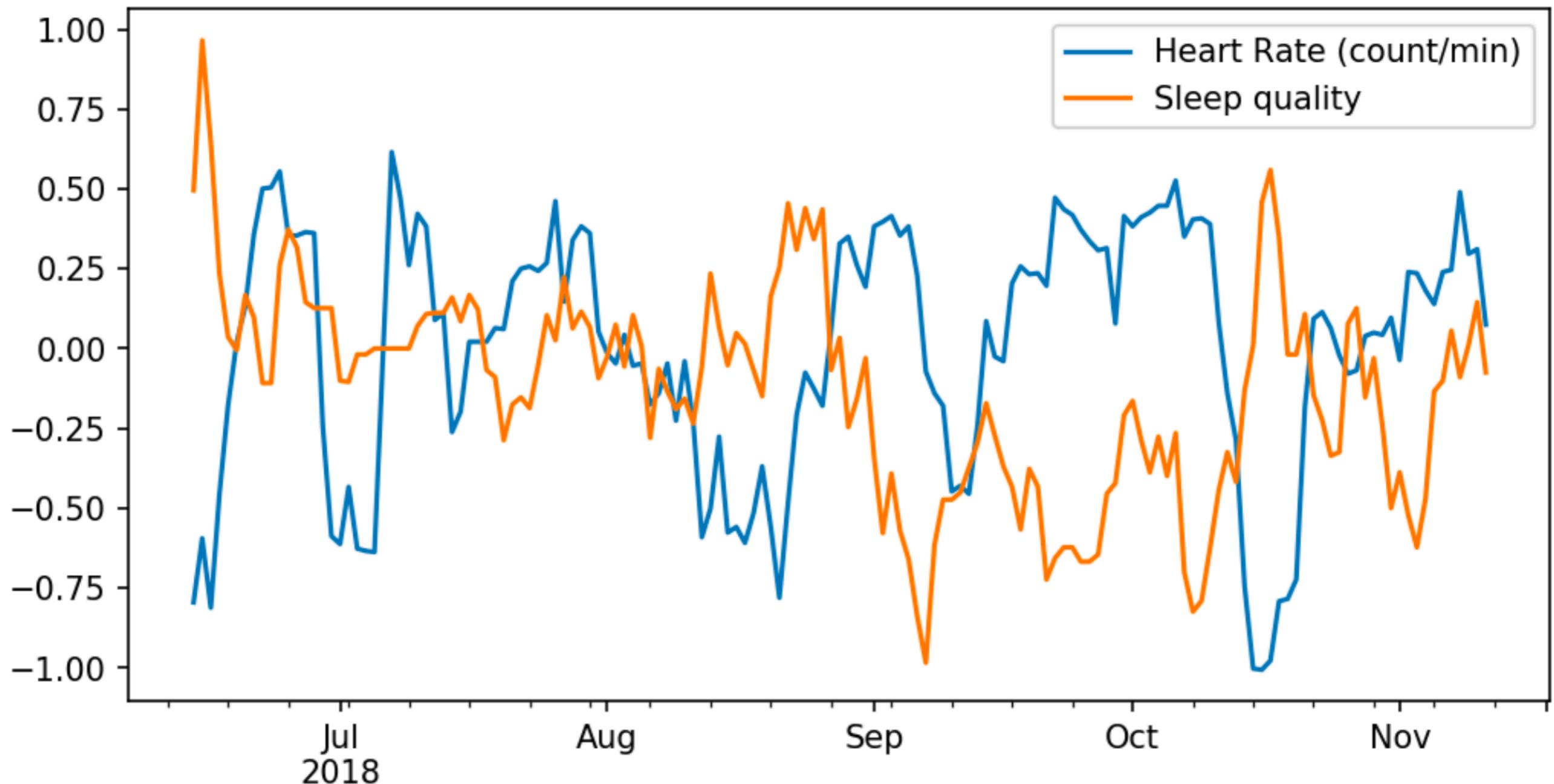
Sleep quality vs cycling distance (7-day mean, normalized)



```
aligned_hr = align_and_compare(normalized_quality, normalized_heartrate)
plt.title('Sleep quality vs heart rate average (7-day mean, normalized)')
plt.legend()
plt.show()
```

r = -0.11030434760685069, p-value = 0.17756284414832138

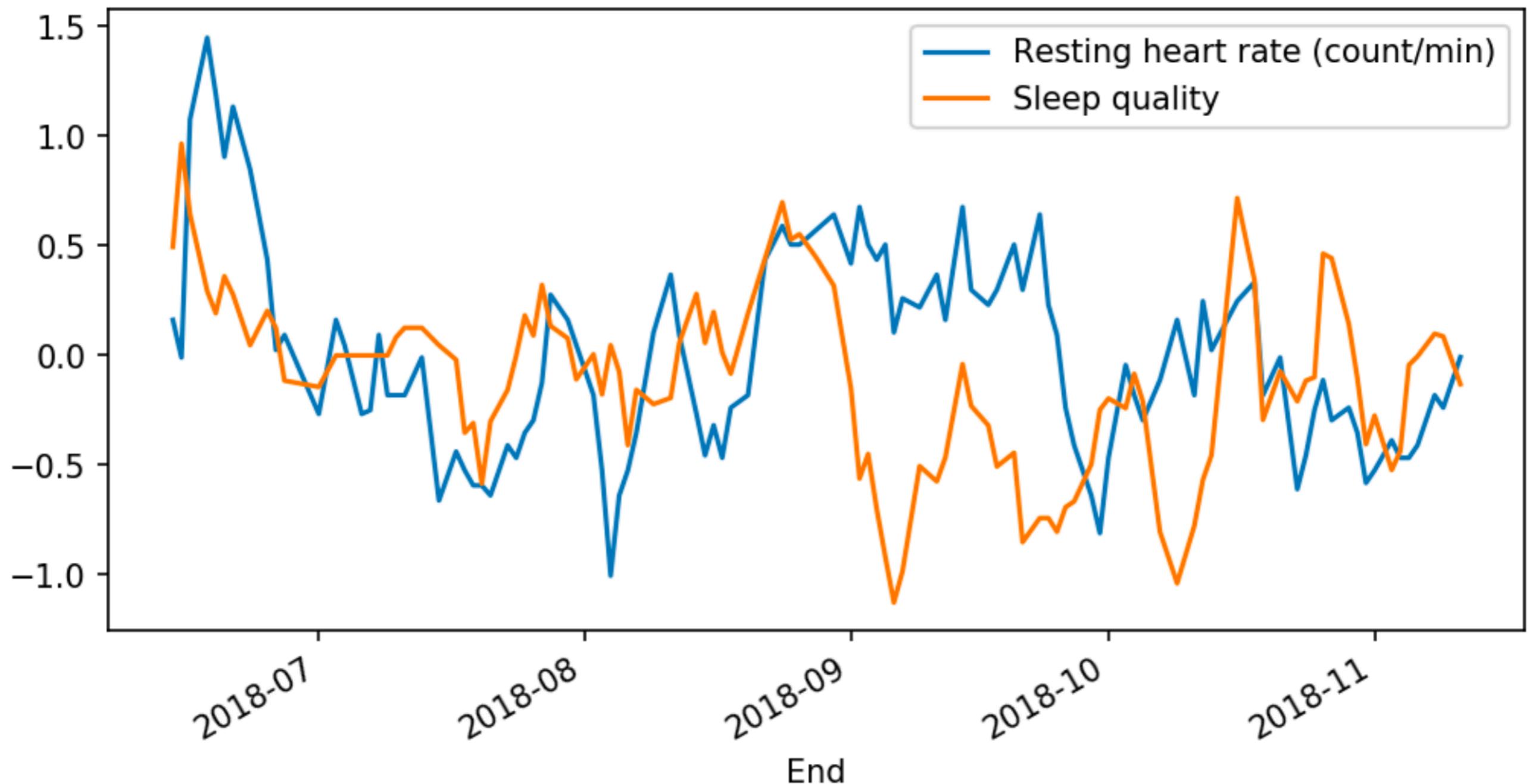
Sleep quality vs heart rate average (7-day mean, normalized)



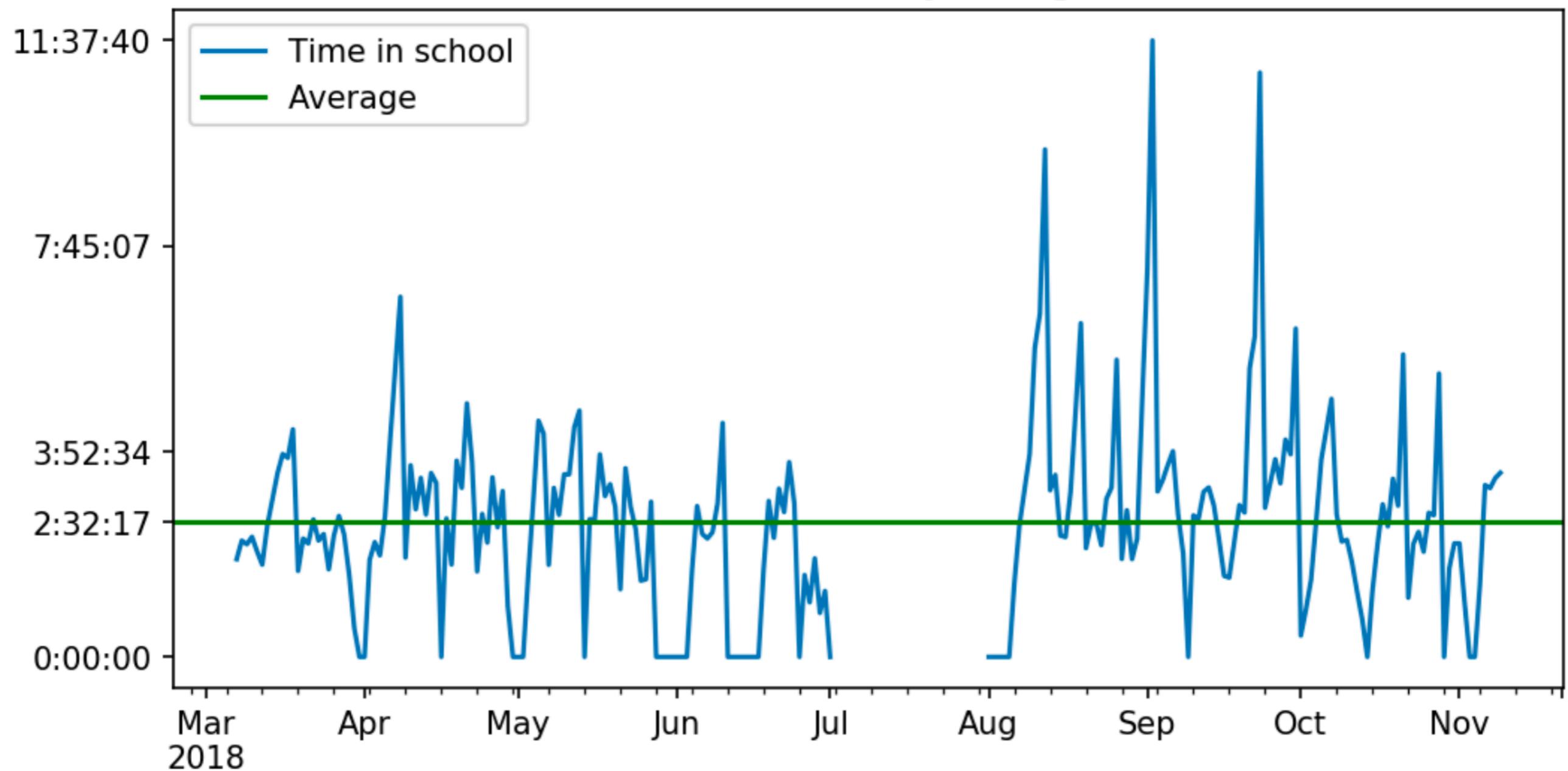
```
aligned_resting = align_and_compare(normalized_quality, normalized_resting)
plt.title('Sleep quality vs resting heart rate average (7-day mean, normalized)')
plt.legend()
plt.show()
```

r = 0.08934727412678754, p-value = 0.35780248929367564

Sleep quality vs resting heart rate average (7-day mean, normalized)



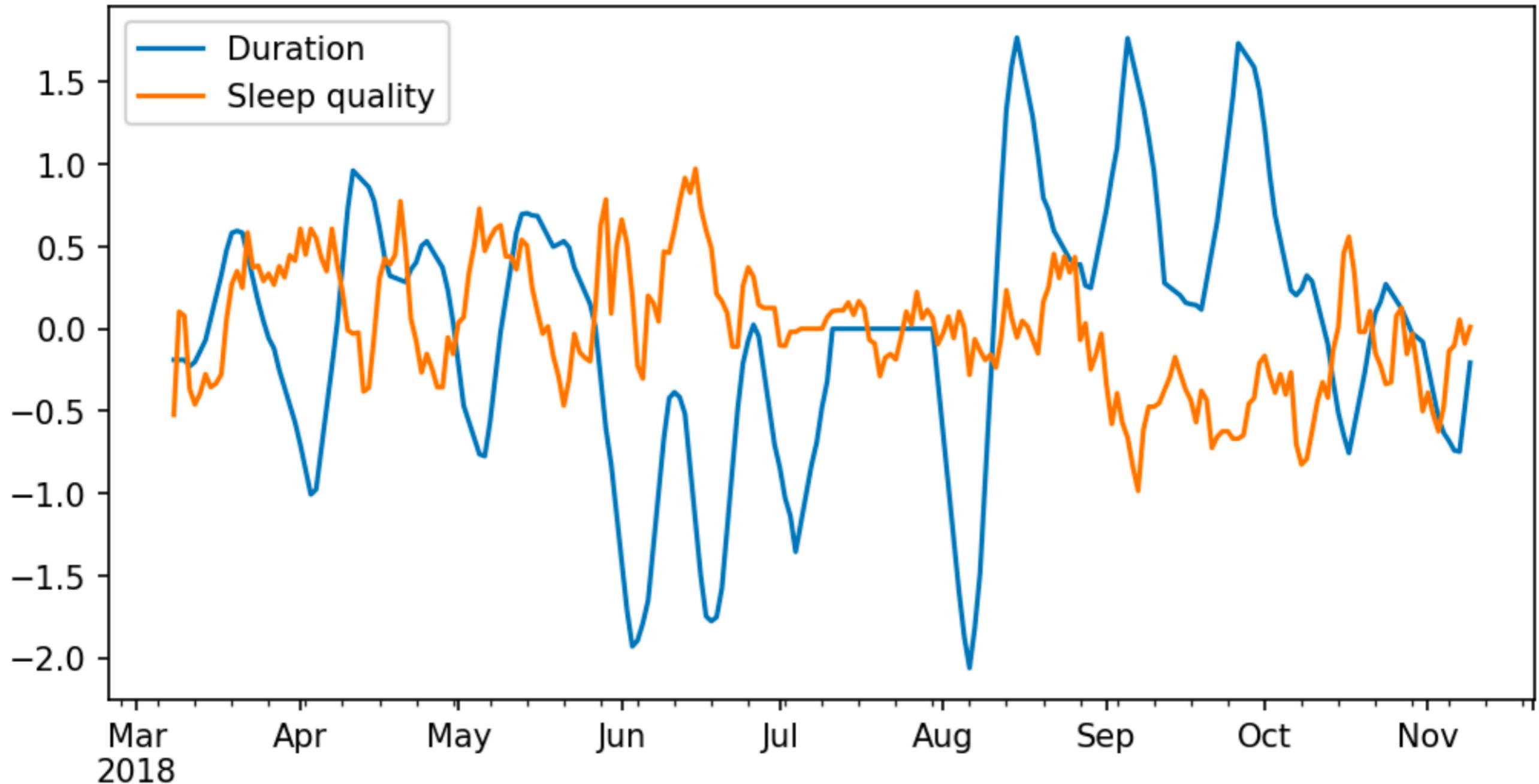
Time in school, 3-day rolling mean



```
normalized_school = normalize_series(school.rolling('7d').mean())
aligned_school = align_and_compare(normalized_quality, normalized_school)
plt.title('Time in school vs sleep quality (7-day mean, normalized)')
plt.legend()
plt.show()
```

r = -0.1789462436620316, p-value = 0.004788912541843903

Time in school vs sleep quality (7-day mean, normalized)



```

data = pandas.concat((sleepquality.fillna(sleepquality.mean()).rename('Quality'),
                     normalized_time.rename('Time'),
                     aligned_lateness,
                     aligned_cycling.rename('Cycle_distance'),
                     aligned_hr.rename('HR'),
                     aligned_school.rename('School')), axis=1)
model = smf.ols('Quality ~ Time + Lateness + Cycle_distance + HR + School', data).fit()
print(model.summary())

```

OLS Regression Results

=====

Dep. Variable:	Quality	R-squared:	0.773
Model:	OLS	Adj. R-squared:	0.763
Method:	Least Squares	F-statistic:	72.38
Date:	Thu, 22 Nov 2018	Prob (F-statistic):	1.37e-32
Time:	19:09:57	Log-Likelihood:	162.15
No. Observations:	112	AIC:	-312.3
Df Residuals:	106	BIC:	-296.0
Df Model:	5		
Covariance Type:	nonrobust		

=====

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.7152	0.007	106.205	0.000	0.702	0.729
Time	0.1128	0.006	18.290	0.000	0.101	0.125
Lateness	0.0010	0.007	0.149	0.882	-0.012	0.014
Cycle_distance	-0.0072	0.007	-1.001	0.319	-0.021	0.007
HR	0.0047	0.008	0.570	0.570	-0.012	0.021
School	-0.0128	0.006	-2.315	0.023	-0.024	-0.002

=====

Omnibus:	3.835	Durbin-Watson:	1.960
Prob(Omnibus):	0.147	Jarque-Bera (JB):	3.938
Skew:	-0.187	Prob(JB):	0.140
Kurtosis:	3.839	Cond. No.	2.32

=====

```
model = smf.ols('Quality ~ Time + School', data).fit()
print(model.summary())
```

OLS Regression Results

```
=====
Dep. Variable:                 Quality      R-squared:                   0.690
Model:                          OLS          Adj. R-squared:             0.687
Method:                         Least Squares   F-statistic:                  227.7
Date:              Thu, 22 Nov 2018   Prob (F-statistic):        8.33e-53
Time:                19:10:01           Log-Likelihood:            255.41
No. Observations:                  208          AIC:                      -504.8
Df Residuals:                      205          BIC:                      -494.8
Df Model:                           2
Covariance Type:                nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.7178	0.005	144.563	0.000	0.708	0.728
Time	0.1035	0.005	20.737	0.000	0.094	0.113
School	-0.0151	0.005	-2.863	0.005	-0.026	-0.005

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Omnibus:	10.647	Durbin-Watson:	1.790
Prob(Omnibus):	0.005	Jarque-Bera (JB):	14.335
Skew:	-0.352	Prob(JB):	0.000771
Kurtosis:	4.076	Cond. No.	1.15

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```

model = smf.ols('Quality ~ Time + Time.shift(-1) + School', data).fit()
print(model.summary())

```

OLS Regression Results

	Quality	R-squared:	0.679
Dep. Variable:	Quality	Model:	OLS
Method:	Least Squares	R-squared:	0.673
Date:	Thu, 22 Nov 2018	Adj. R-squared:	0.673
Time:	19:22:48	F-statistic:	131.0
No. Observations:	190	Prob (F-statistic):	1.26e-45
Df Residuals:	186	Log-Likelihood:	230.49
Df Model:	3	AIC:	-453.0
Covariance Type:	nonrobust	BIC:	-440.0

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.7178	0.005	135.113	0.000	0.707	0.728
Time	0.1020	0.005	19.221	0.000	0.092	0.112
Time.shift(-1)	-0.0043	0.006	-0.763	0.446	-0.015	0.007
School	-0.0149	0.006	-2.601	0.010	-0.026	-0.004

Omnibus:	7.750	Durbin-Watson:	1.797
Prob(Omnibus):	0.021	Jarque-Bera (JB):	9.488
Skew:	-0.301	Prob(JB):	0.00870
Kurtosis:	3.915	Cond. No.	1.20