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dr. Moira van Leeuwen

Foot Placement Control & Corticomuscular Coherence

Concepts &
Analysis Pipeline

Amsterdam
Movement
Sciences

VU VRIJE
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 Amsterdam UMC
University Medical Centers

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Foot Placement Control

&
Corticomuscular Coherence

Concepts &
Analysis Pipeline

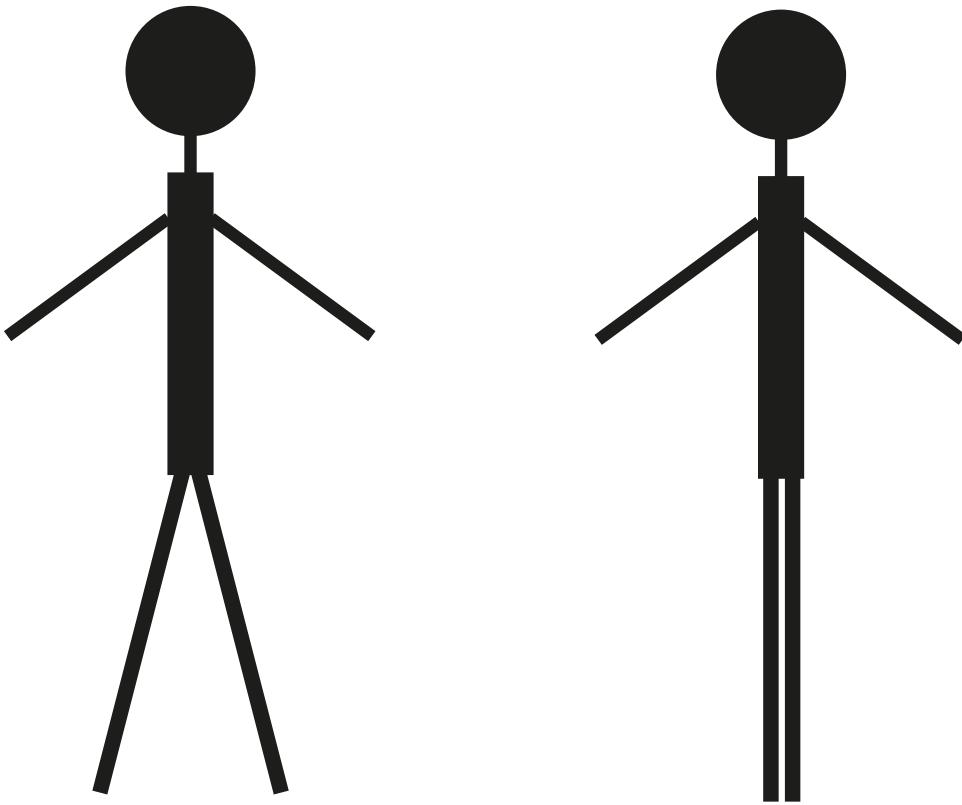
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Stability control

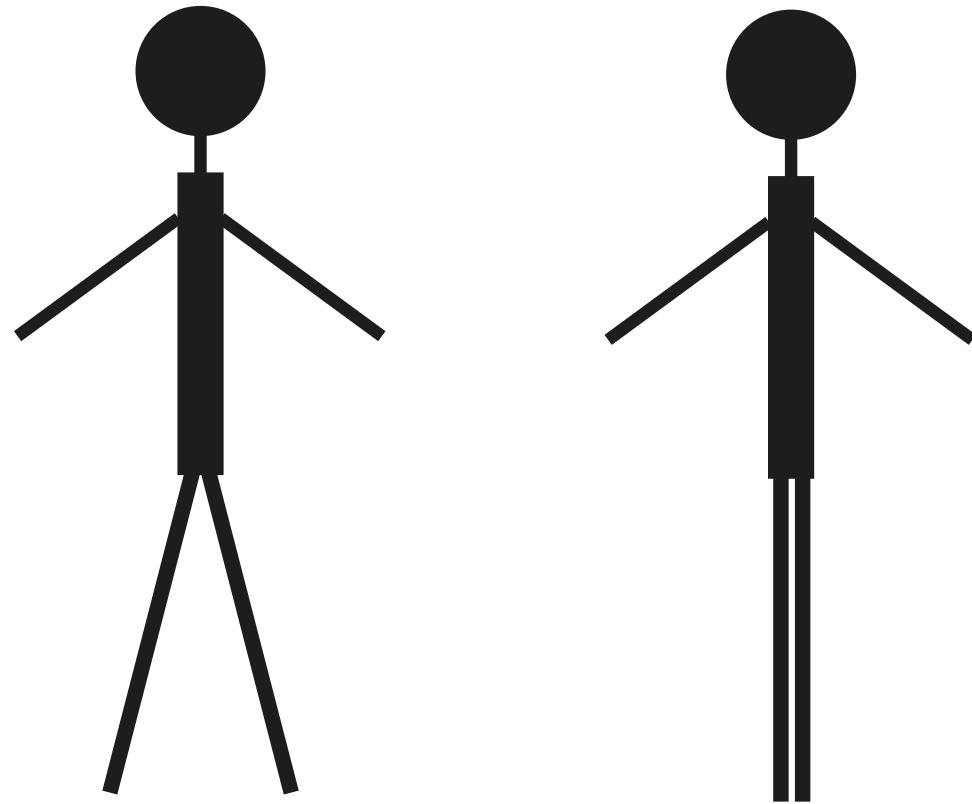
Coordinating the base
of support wrt the
center of mass



Stability control

Coordinating the base
of support wrt the
center of mass

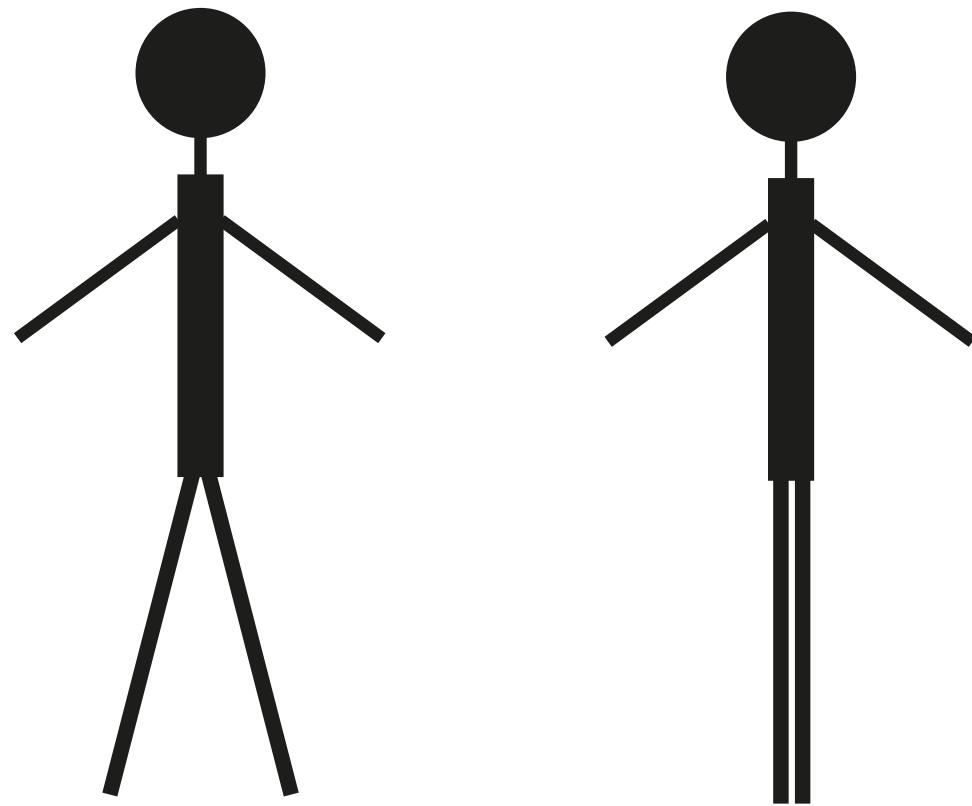
Center-of-mass



Stability control

Coordinating the base
of support wrt the
center of mass

Center-of-mass

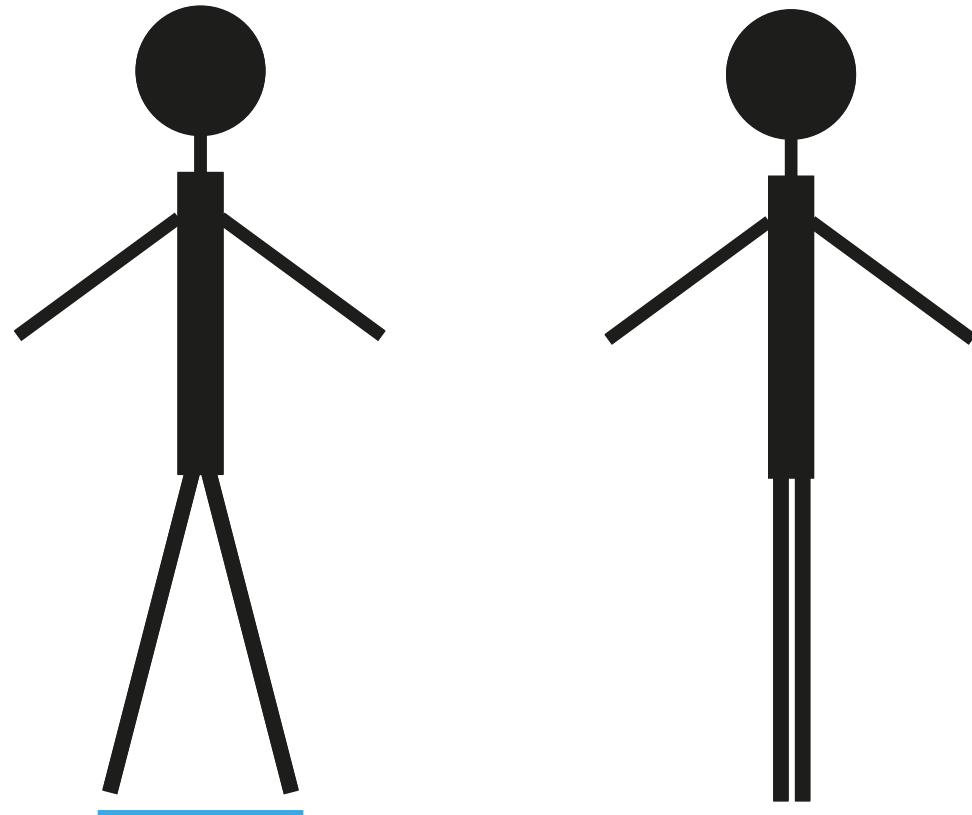


Base of Support

Stability control

Coordinating the base
of support wrt the
center of mass

Center-of-mass

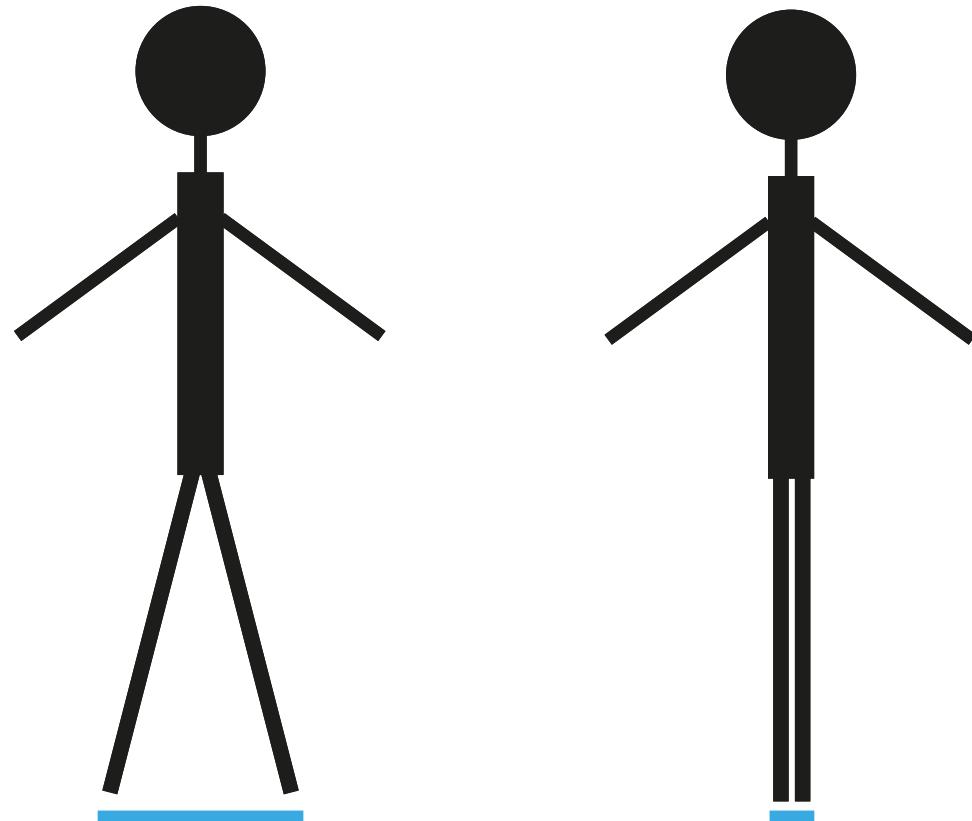


Base of Support

Stability control

Coordinating the base
of support wrt the
center of mass

Center-of-mass

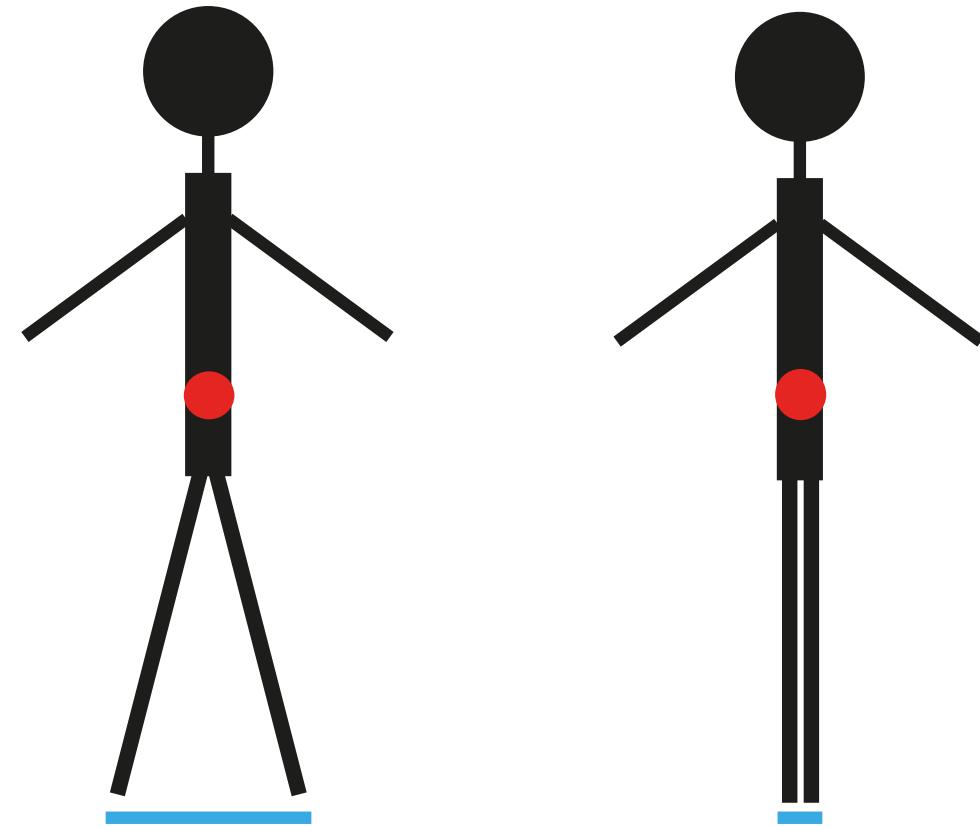


Base of Support

Stability control

Coordinating the base
of support wrt the
center of mass

Center-of-mass

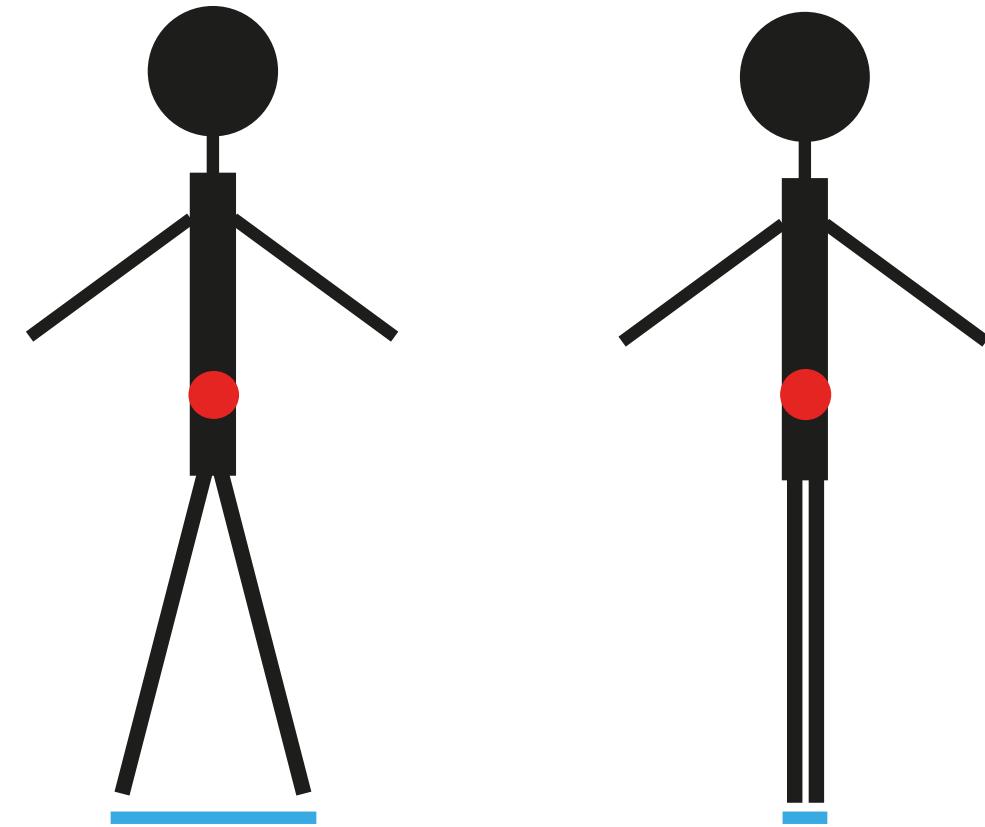


Base of Support

Stability control

Coordinating the base
of support wrt the
center of mass

Center-of-mass

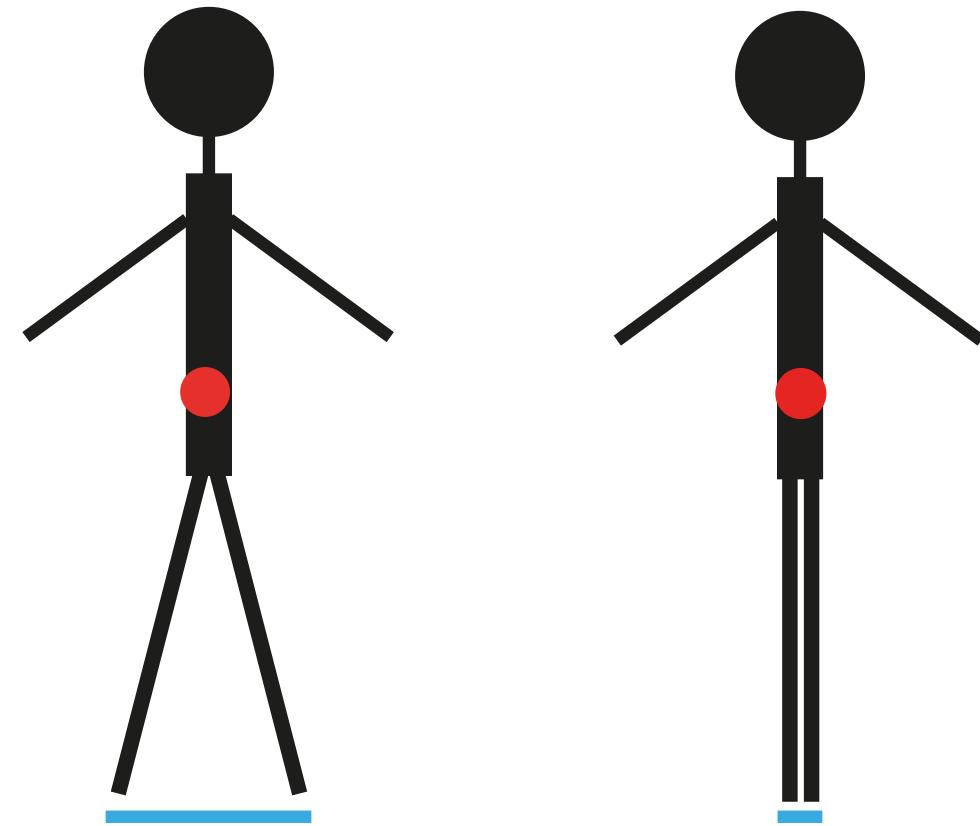


Base of Support

Stability control

Coordinating the base
of support wrt the
center of mass

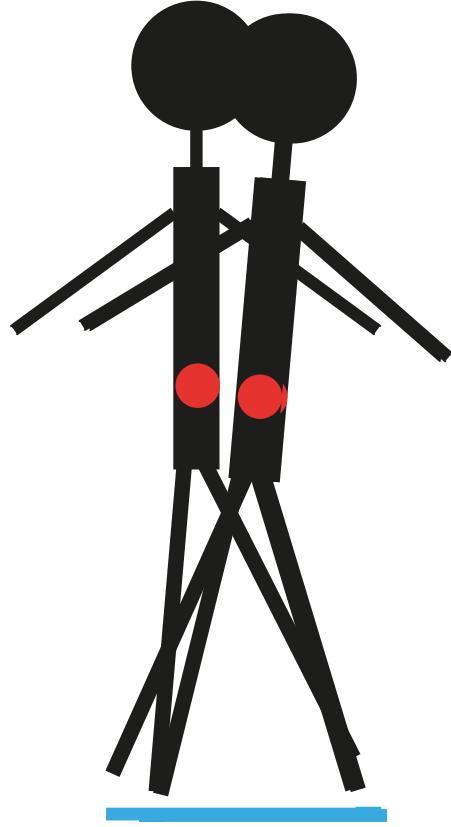
Center-of-mass



Base of Support

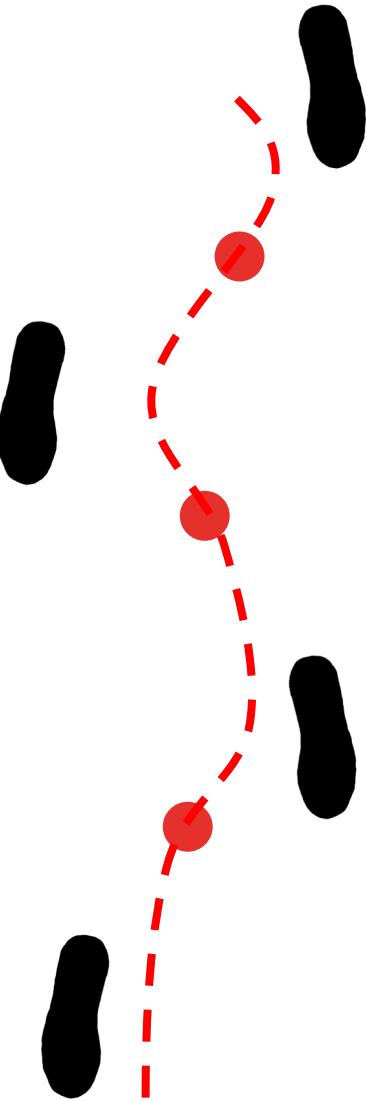
Stability control

Foot placement is dominant stability mechanism during walking



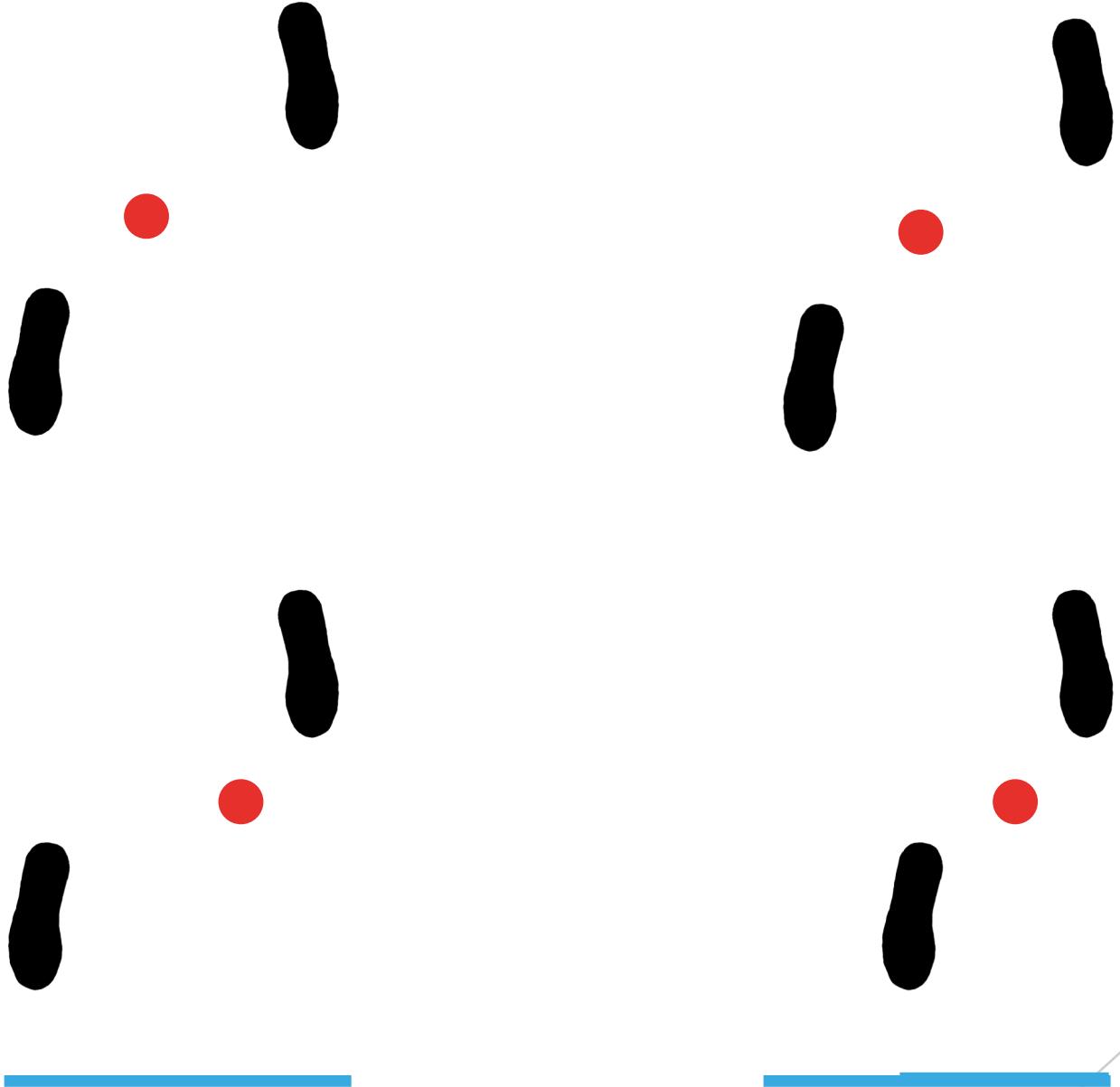
Stability control

Foot placement is dominant stability mechanism during walking



Stability control

Foot placement is dominant stability mechanism during walking



$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

Stability control

Foot placement control can be quantified using a linear feedback model

FP

Step Width

Step Length

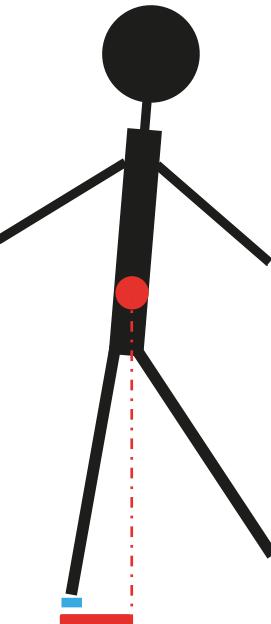


Stability control

Foot placement control can be quantified using a linear feedback model

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

CoM_{pos}
 CoM_{vel}



**ML/AP
CoM position**

Wrt stance leg

Derivative

Stability control

Foot placement control can be quantified using a linear feedback model

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

β
 β

**Regression coefficient
Gain
Strength of FP response**

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

Stability control

Foot placement control can be quantified using a linear feedback model

i

Percentage of swing phase

Stability control

Foot placement control can be quantified using a linear feedback model

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

ε

Residual

Foot placement error

Motor noise/loose control

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

Stability control

Foot placement control can be quantified using a linear feedback model



```
stats = regstats(foot_combined,pred_combined(:,1:order), 'linear');
```

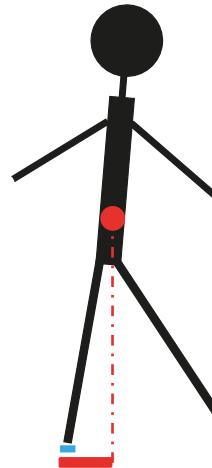
Predictors & Dependent variable

`pred_combined`

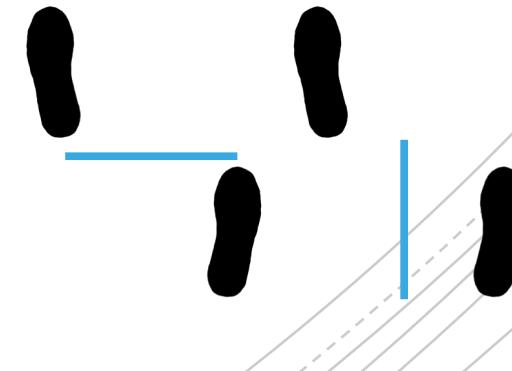
`foot_combined`

```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

- Pelvis marker
- Heel markers

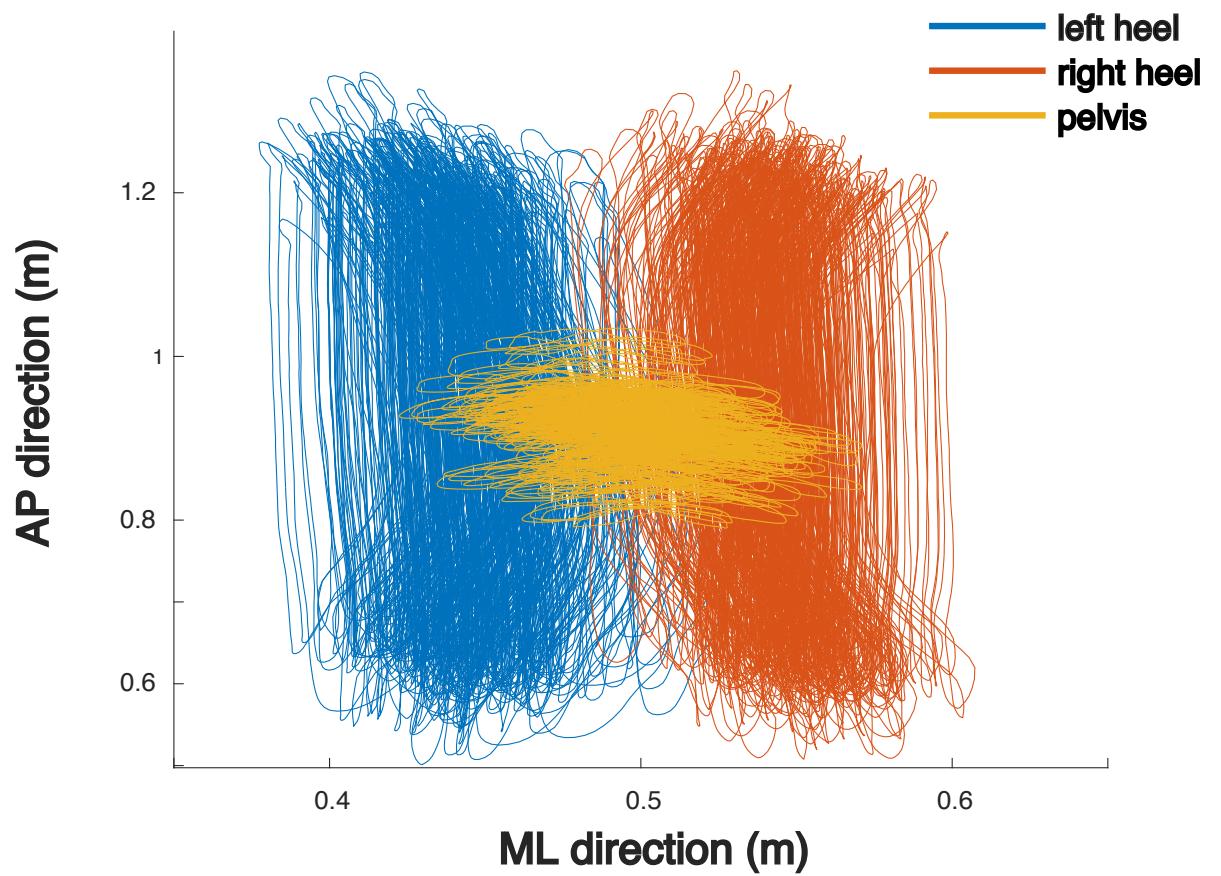


- Heel markers



Pelvis and heel markers

Quick check:
kinematic data



Predictors & Dependent variable

pred_combined

foot_combined

Gait event detection

- Heelstrike & toe-off events
- Heel markers

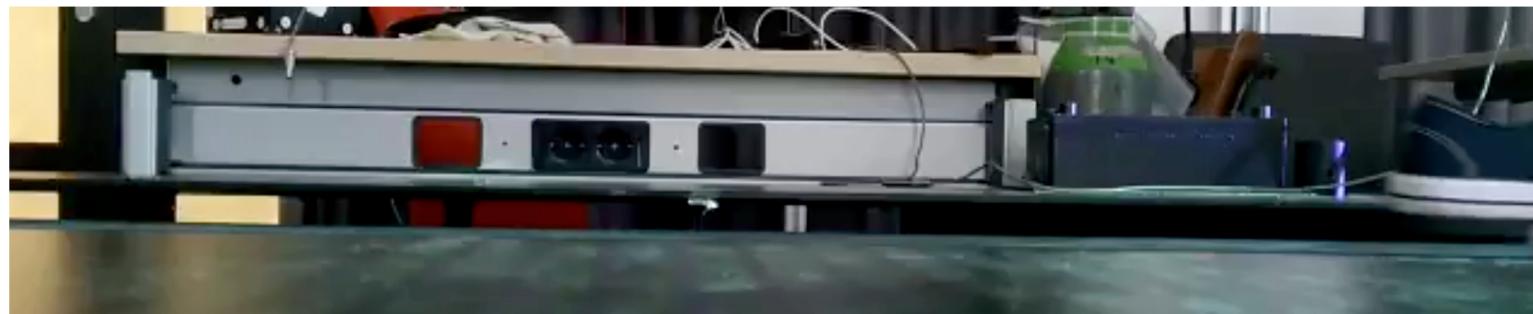
Gait event detection

Detecting events
based on AP velocity

Zeni et al.(2008)

Anterior posterior velocity heelmarker

- AP velocity + → - : Heelstrike
- AP velocity - → + : Toe-off



Gait event detection

Detecting events
based on AP velocity

Zeni et al.(2008)

`[lhs,rto,rhs,lto] = calc_events_zeni(L_heel_AP,R_heel_AP,fs,fs_t)`

- `L_heel_AP` : left heelmarker position data AP direction
- `R_heel_AP` : right heelmarker position data AP direction
- `fs`: sampling frequency motion capture system
- `fs_t`: threshold to avoid false heelstrike detection within `fs_t` seconds of a detected heelstrike

LHS RTO RHS LTO

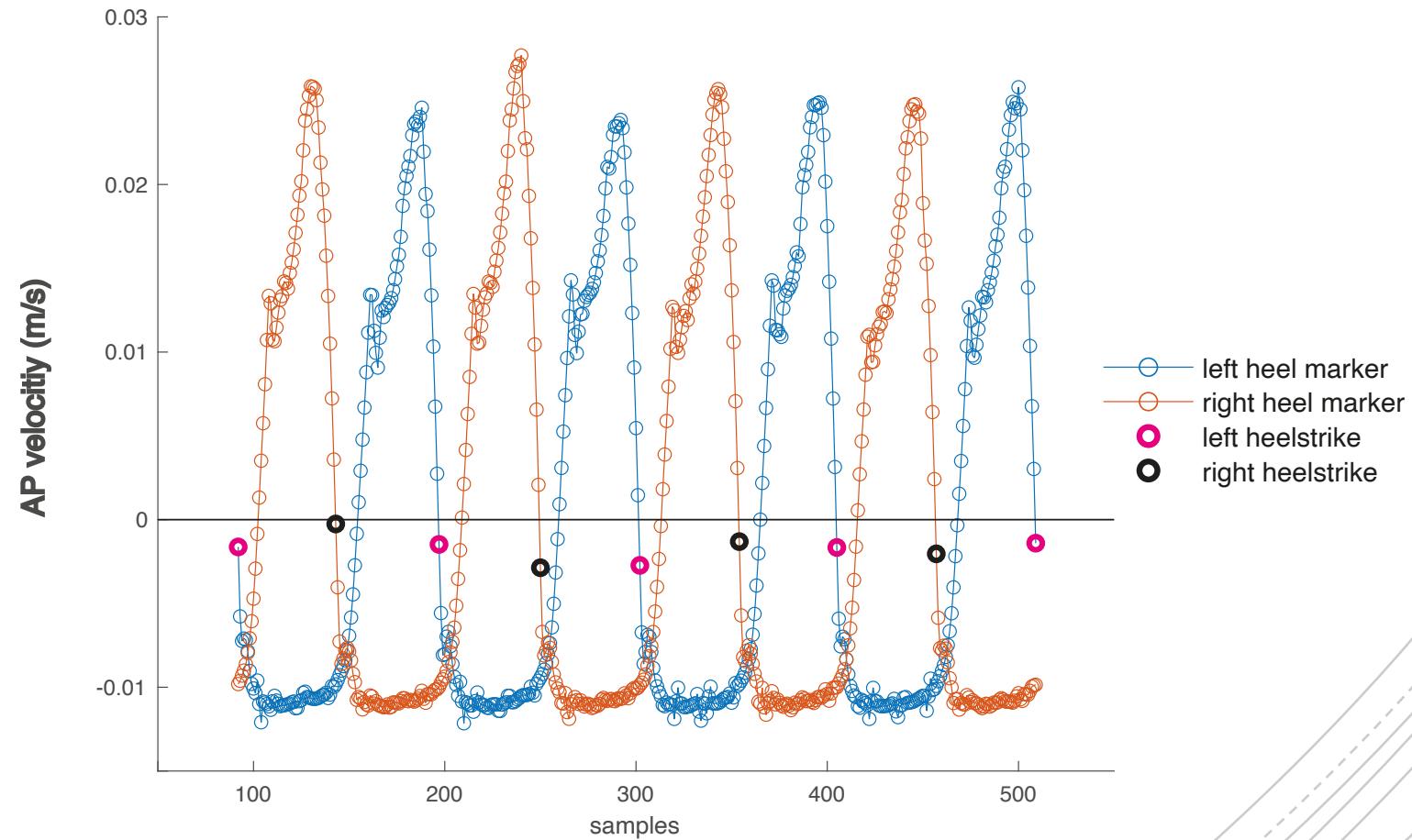
Gait event detection

Detecting events
based on AP velocity

Zeni et a.(2008)

```
[lhs,rto,rhs,lto] = calc_events_zeni(L_heel_AP,R_heel_AP,fs,fs_t)
```

Zeni heelstrike detection (8 steps)



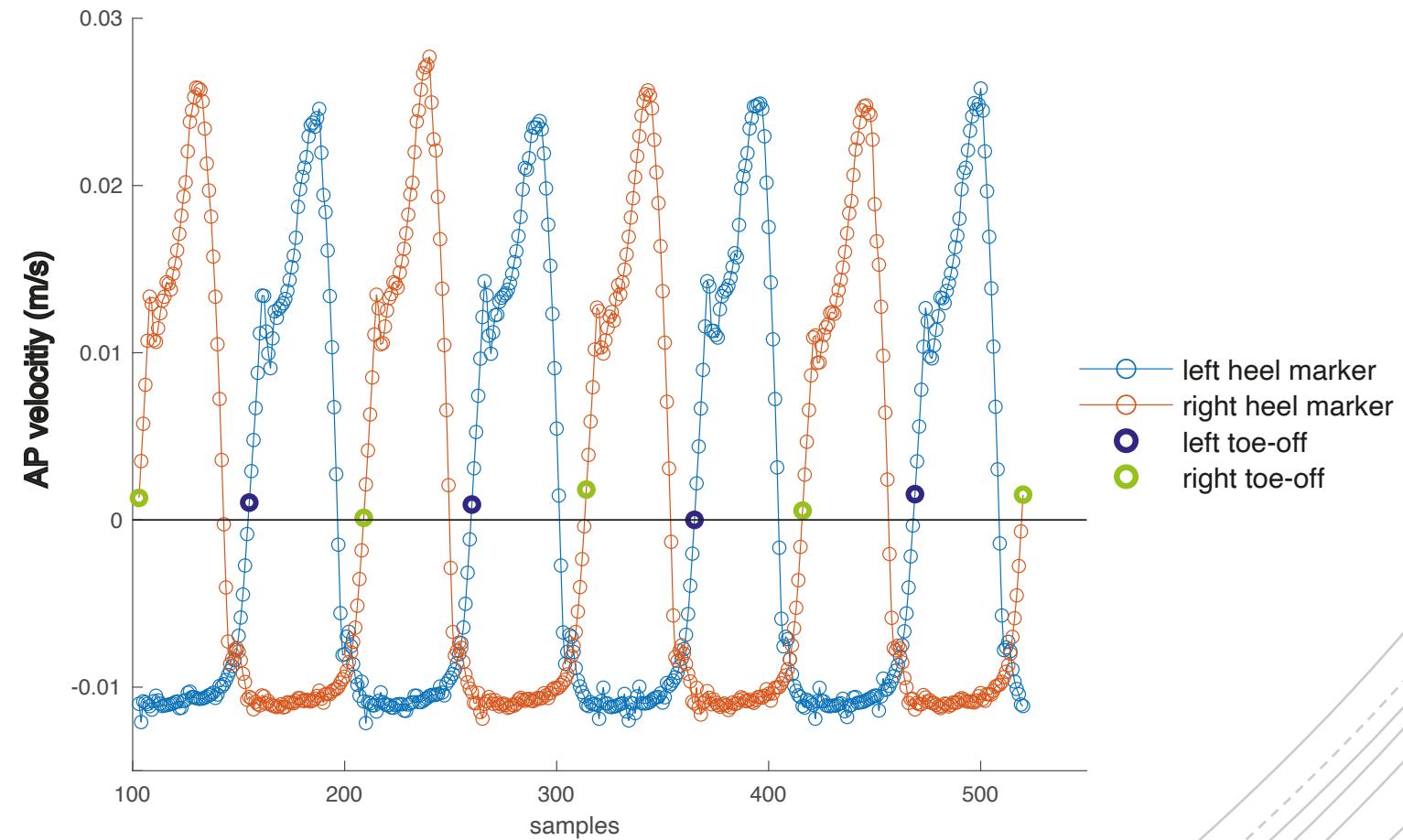
Gait event detection

Detecting events
based on AP velocity

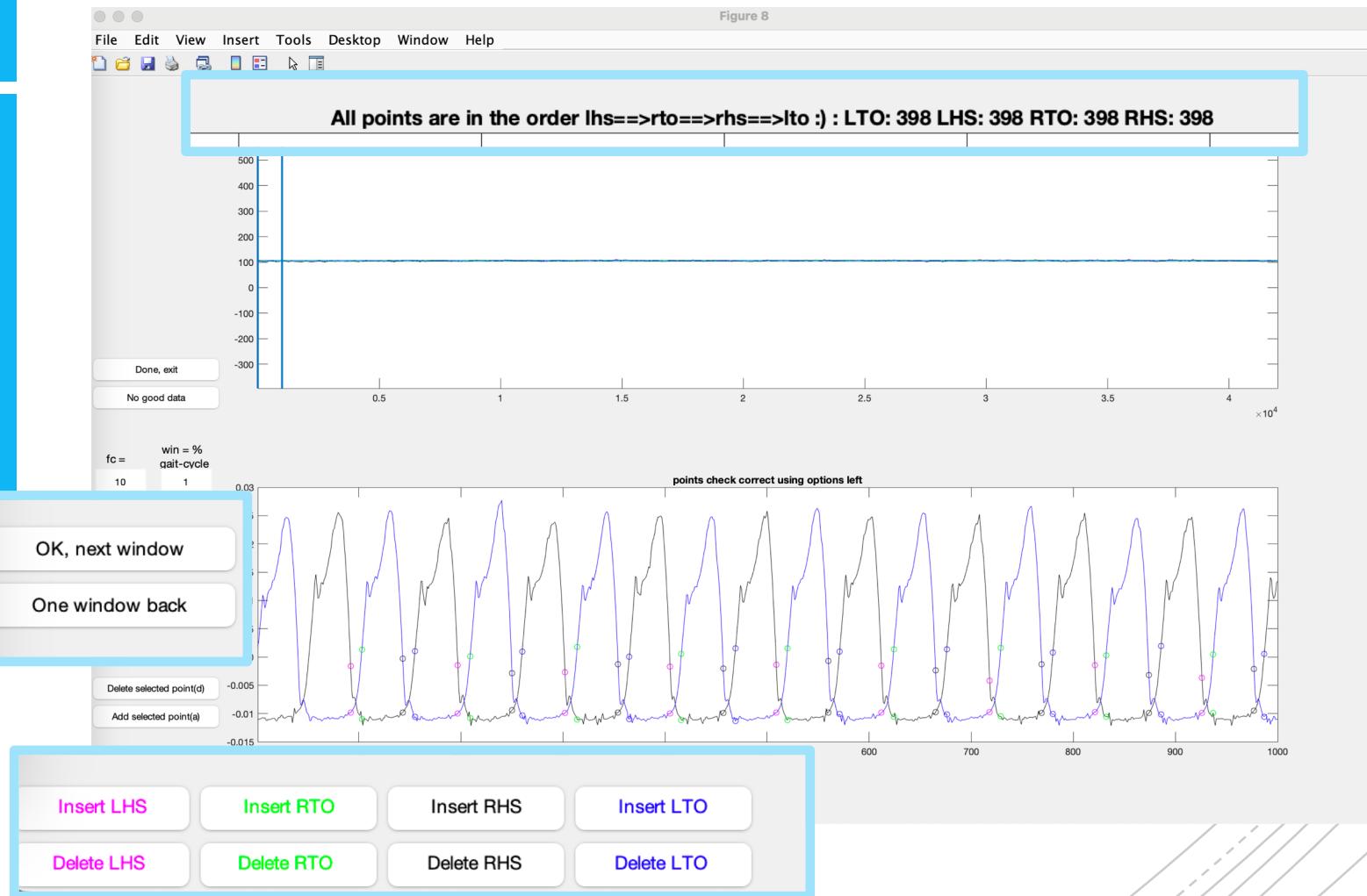
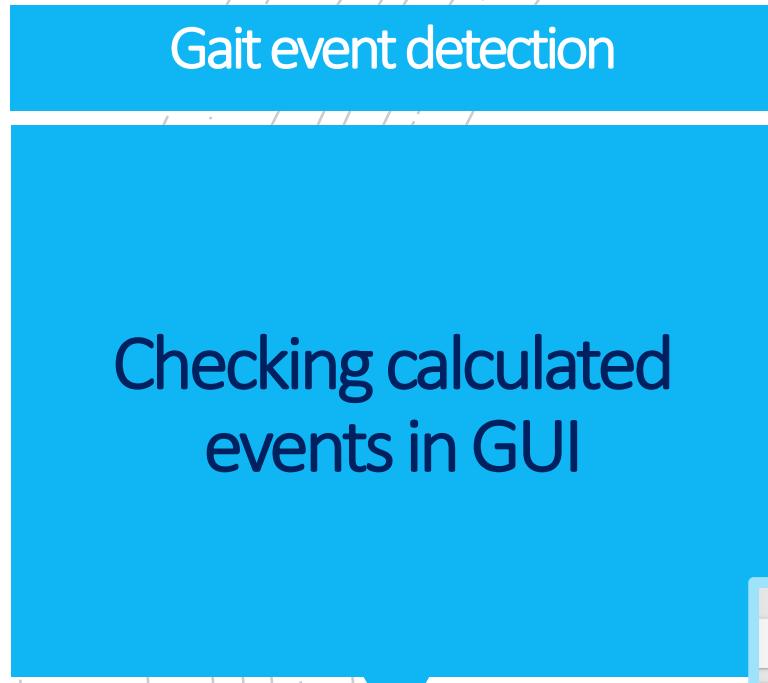
Zeni et a.(2008)

```
[lhs,rto rhs,lto] = calc_events_zeni(L_heel_AP,R_heel_AP,fs,fs_t)
```

Zeni toe-off detection (8 steps)



>  =VU 3D model=



[events,notgood]=check_signal(events, signal, fs)

Gait event detection

Checking gait events in GUI

The image displays three screenshots of a gait analysis software interface. The top screenshot shows a 3D skeleton walking on a ground plane, with a butterfly-shaped diagram indicating the Center of Pressure (Anterior) and four circular markers labeled RTO (Right Toe Off), LTO (Left Toe Off), RHS (Right Heel Strike), and LHS (Left Heel Strike). The middle screenshot is a video player showing a graph of Center of Pressure over time, with a red arrow pointing to a blue downward spike indicating an additional Left Toe Off (LTO). The bottom screenshot shows a detailed view of a graph with multiple colored lines and a keyboard overlay showing shortcut keys for adding and deleting events.

Gait Events Detection Using Center of Pressure

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OK, next window

One window back

Delete selected point(d)

Add selected point(a)

Insert LHS Insert RTO Insert RHS

Delete LHS Delete RTO Delete RHS

Our primary objectives

- 1) Sequencing all events as LHS==>RHS==>LTO
- 2) Ensuring equal occurrence counts for all events.

Shortcut keys

For adding a missing event

For deleting an extra event

Zoom in

Zoom out

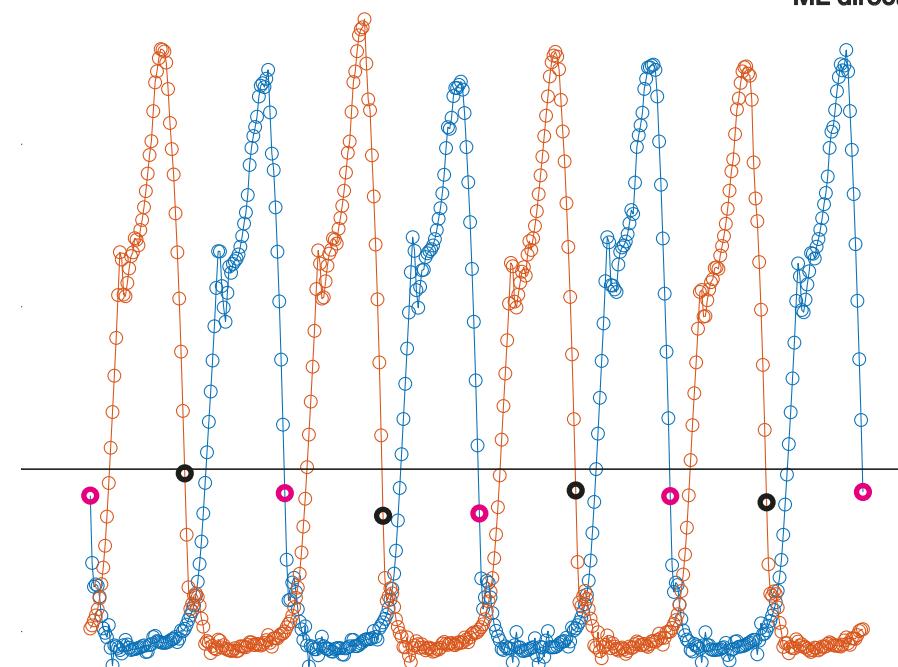
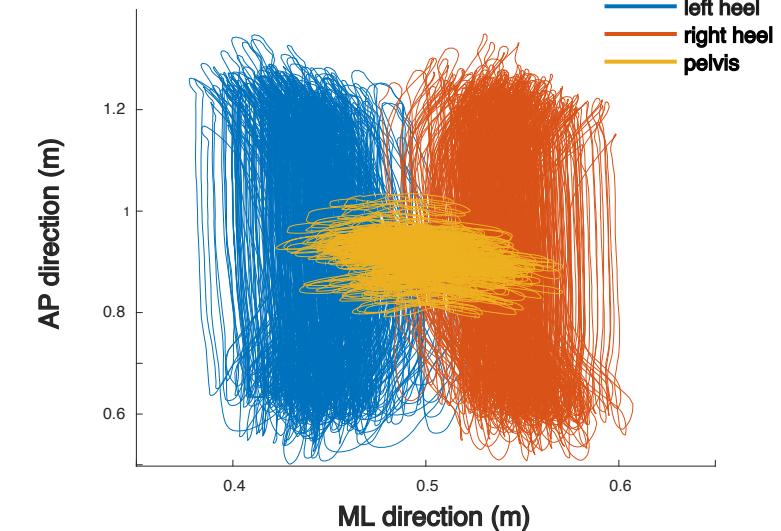
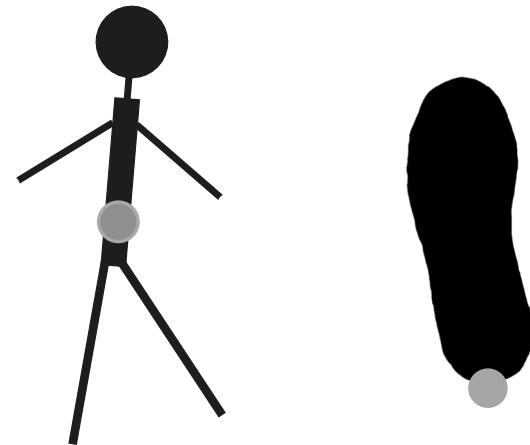
Navigate left right

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Kinematics_Gait Event Detection

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```
[OUT,intermediates] =  
foot_placement_model_function_step(CoM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
stats = regstats(foot_combined,pred_combined(:,1:order),'linear');
```



```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

COM

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Rfoot Lfoot

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

events

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

fsopto

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

pred_samples
1:51

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

order

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

removeorigin

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

centerdata

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

- Taking derivatives
- Time normalization
- Removing origin
- Centering the data
- Combining Left and Right steps

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Taking derivatives

Foot placement model

Function preparing
foot_combined
&
pred_combined

COM fsopto

```
COM_vel = calc_derivative(COM,fsopto);  
COM_acc = calc_derivative(COM_vel,fsopto);
```

order

- 1 → CoM position only
- 2 → CoM position and velocity
- 3 → CoM position and velocity and acceleration

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Time normalization

Foot placement model

Function preparing
foot_combined
&
pred_combined

COM Rfoot Lfoot events

```
COM_L = normalizetimebase_step(COM,lto(1:end-1),lhs(2:end));  
COM_L_vel = normalizetimebase_step(COM Vel,lto(1:end-1),lhs(2:end));
```

Swing phase preceding foot placement

```
foot_L = normalizetimebase_step(Lfoot,rto(2:end),rhs(2:end));
```

Single stance phase of the placed foot

```
origin_L = normalizetimebase_step(Rfoot,lto(1:end-1),lhs(2:end));
```

Single stance phase of stance foot wrt the other foot is placed to

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Removing origin

(... with respect to?)

removeorigin

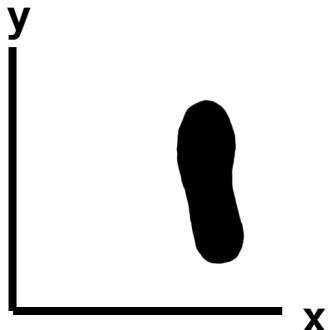
```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Removing origin (... with respect to?)

removeorigin RfootLfoot



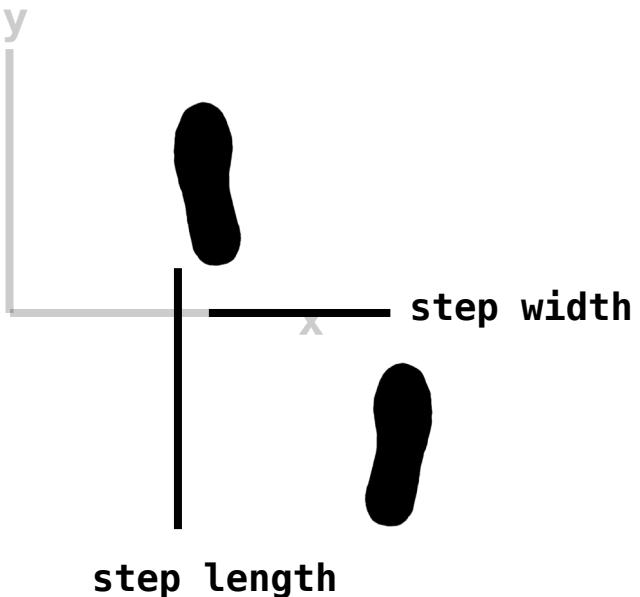
```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Removing origin (... with respect to?)

removeorigin Rfoot Lfoot



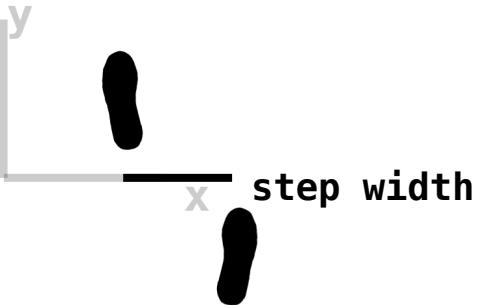
```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Removing origin (... with respect to?) Mediolateral

removeorigin



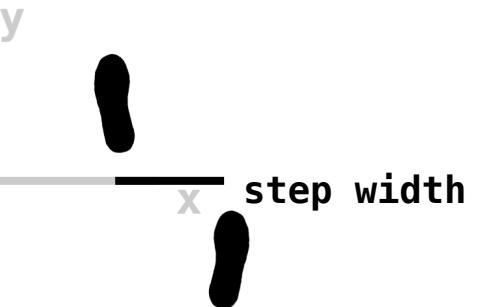
```
foot_L = foot_L(25,:');  
origin_L = origin_L(25,:');  
foot_R = foot_R(25,:');  
origin_R = origin_R(25,:');
```

```
if removeorigin  
foot_L = foot_L -origin_L;  
foot_R = foot_R -origin_R ;  
end
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined



Removing origin (... with respect to?) Mediolateral midstance

removeorigin Rfoot Lfoot

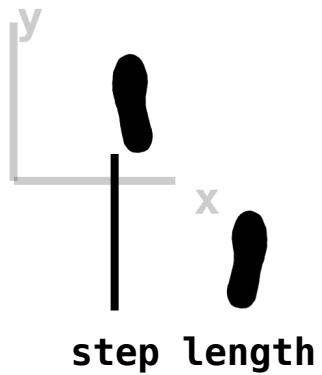
```
foot_L = foot_L(25,:');  
origin_L = origin_L(25,:');  
foot_R = foot_R(25,:');  
origin_R = origin_R(25,:');
```

```
if removeorigin  
foot_L = foot_L -origin_L;  
foot_R = foot_R -origin_R ;  
end
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined



Removing origin (... with respect to?) Anteroposterior

removeorigin Rfoot Lfoot

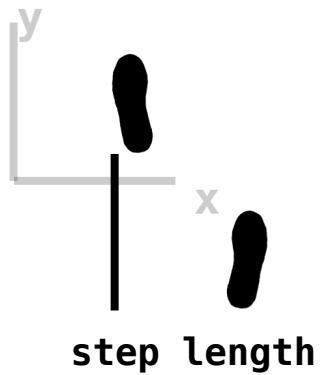
```
foot_L = Lfoot(lhs(2:end));  
origin_L = origin_L(end,:>');  
foot_R = Rfoot(rhs(1:end-1));  
origin_R = origin_R(end,:>');
```

```
if removeorigin  
foot_L = foot_L -origin_L;  
foot_R = foot_R -origin_R ;  
end
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined



Removing origin (... with respect to?) Anteroposterior heelstrike

removeorigin Rfoot Lfoot

```
foot_L = Lfoot(lhs(2:end));  
origin_L = origin_L(end,:>');  
foot_R = Rfoot(rhs(1:end-1));  
origin_R = origin_R(end,:>');
```

```
if removeorigin  
foot_L = foot_L -origin_L;  
foot_R = foot_R -origin_R ;  
end
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Centering the data (subtracting the mean)

centerdata

```
if centerdata  
foot_L = foot_L -nanmean(foot_L);  
foot_R = foot_R -nanmean(foot_R);  
end
```

if centerdata

```
pred_Lstance = pred_Lstance-  
repmat(nanmean(pred_Lstance),size(pred_Lstance,1),1);  
pred_Rstance = pred_Rstance-  
repmat(nanmean(pred_Rstance),size(pred_Rstance,1),1);  
end
```

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
foot_combined
&
pred_combined

Centering the data (subtracting the mean)

centerdata

```
if centerdata  
foot_L = foot_L -nanmean(foot_L);  
foot_R = foot_R -nanmean(foot_R);  
end
```

if centerdata

```
pred_Lstance = pred_Lstance-  
repmat(nanmean(pred_Lstance),size(pred_Lstance,1),1);  
pred_Rstance = pred_Rstance-  
repmat(nanmean(pred_Rstance),size(pred_Rstance,1),1);  
end
```

$$FP = \text{mean_step_width} + \beta_{\text{CoM_pos}}(i) * \text{CoM}_{\text{pos}}(i) + \beta_{\text{CoM_vel}}(i) * \text{CoM}_{\text{vel}}(i) + \varepsilon$$

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Function preparing
`foot_combined`
&
`pred_combined`

Combining Left & Right

Mediolateral

```
foot_combined = [foot_R_sample ; - foot_L_sample];  
pred_combined = [pred_Rstance(:,1:order); - pred_Lstance(:,1:order)];
```

Anteroposterior

```
foot_combined = [foot_R_sample ; foot_L_sample];  
pred_combined = [pred_Rstance(:,1:order); pred_Lstance(:,1:order)];
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Combining Left & Right

Foot placement model

Function preparing
foot_combined
&
pred_combined

Mediolateral

```
foot_combined = [foot_R_sample ; -foot_L_sample];  
pred_combined = [pred_Rstance(:,1:order); -pred_Lstance(:,1:order)];
```

Anteroposterior

```
foot_combined = [foot_R_sample ; foot_L_sample];  
pred_combined = [pred_Rstance(:,1:order); pred_Lstance(:,1:order)];
```

Foot placement model

```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');  
  
FP = βCoM_pos(i) * CoMpos(i) + βCoM_vel(i) * CoMvel(i) + ε
```

Outcome measures

Foot placement model

Outcome measures

```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

$$R^2 = 1 - \frac{\sum(\text{actual FP} - \text{predicted FP})^2}{\sum(\text{actual FP} - \text{mean FP})^2}$$

Foot placement model

Outcome measures

```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

unexplained FP variance

$$R^2 = 1 - \frac{\text{unexplained FP variance}}{\text{total FP variance}}$$

Foot placement model

Outcome measures

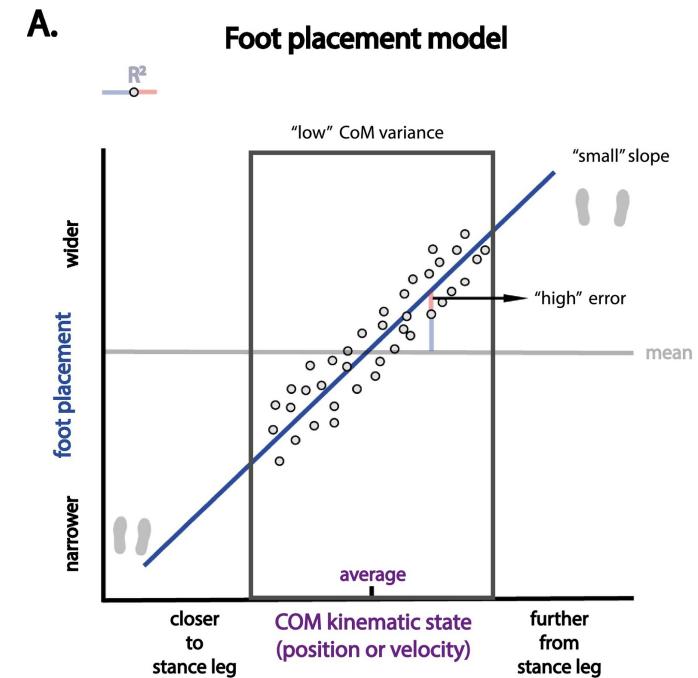
```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

R^2

$$R^2 = 1 - \frac{\text{unexplained FP variance}}{\text{total FP variance}}$$



Foot placement model

Outcome measures

```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

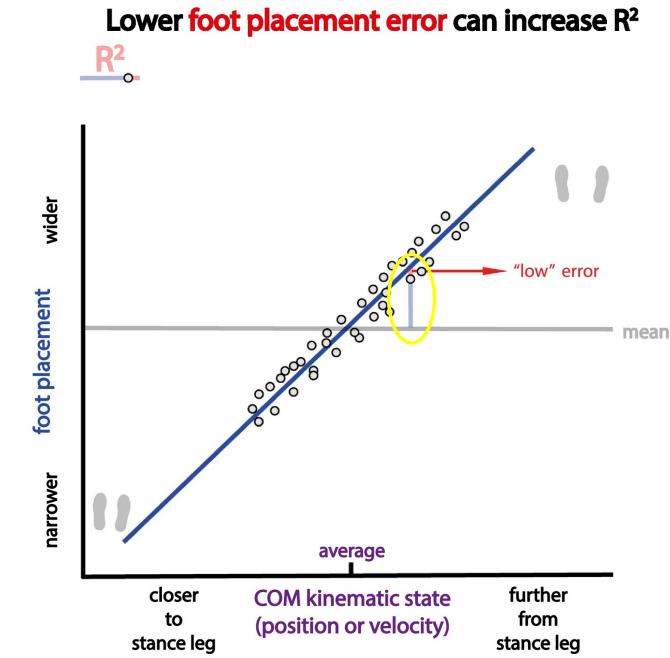
$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

```
intermediates.error_combined(i_pred_sample,ind_combined) = stats.r;
```

The diagram shows two horizontal bars representing the total variance of foot placement. The left bar is divided into two segments: a blue segment labeled R^2 and a red segment labeled "unexplained FP variance". The right bar is also divided into two segments: a blue segment labeled R^2 and a red segment labeled "total FP variance".

$$R^2 = 1 - \frac{\text{unexplained FP variance}}{\text{total FP variance}}$$



Foot placement model

Outcome measures

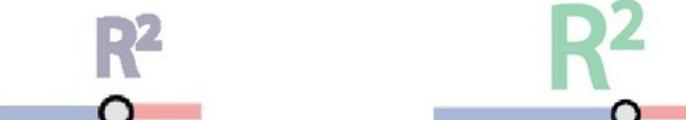
```
stats = regstatsfoot_combined,pred_combined(:,1:order),'linear');
```

$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \epsilon$$

```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

```
OUT.Combined_coeff1.data(i_pred_sample) = stats.beta(2);
```

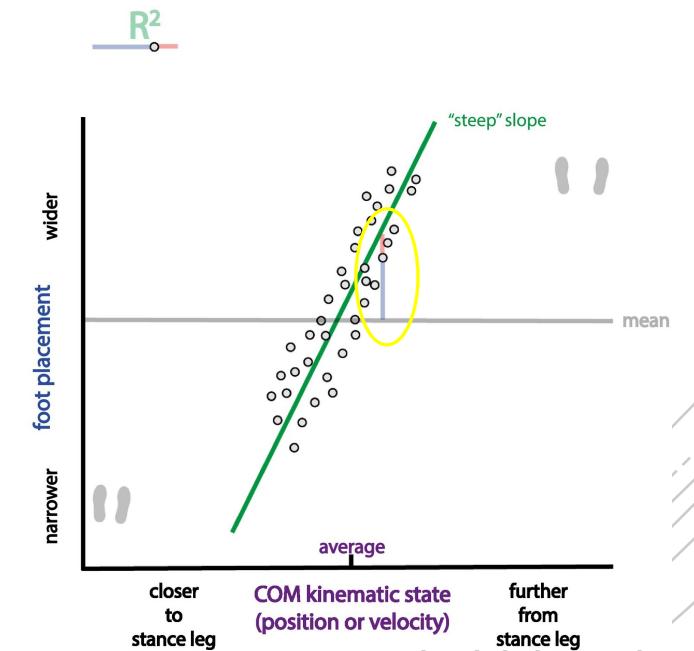
```
OUT.Combined_coeff1.data(i_pred_sample) = stats.beta(3);
```



unexplained FP variance

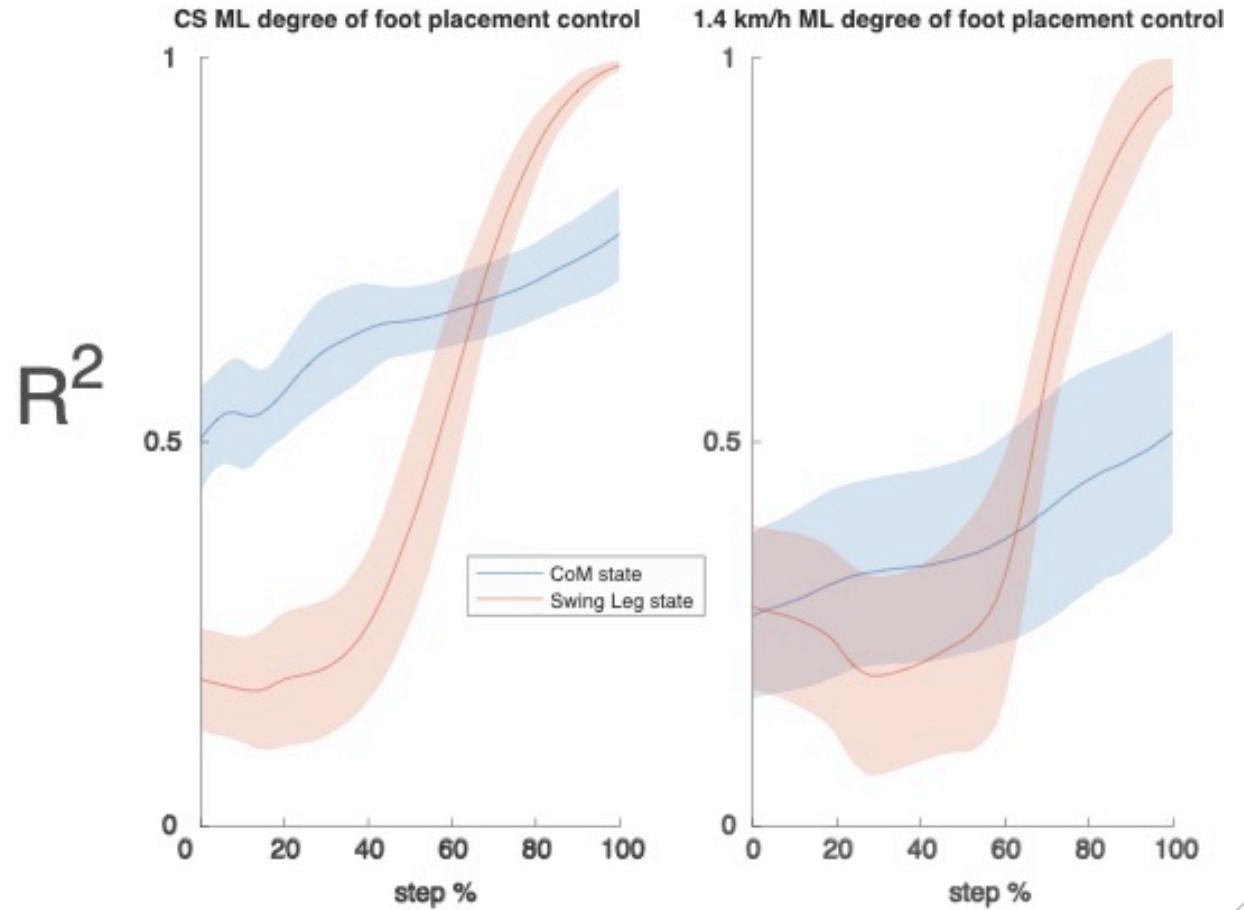
$$R^2 = 1 - \frac{\text{total FP variance}}{n}$$

Higher gain (β) can increase R^2



```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```



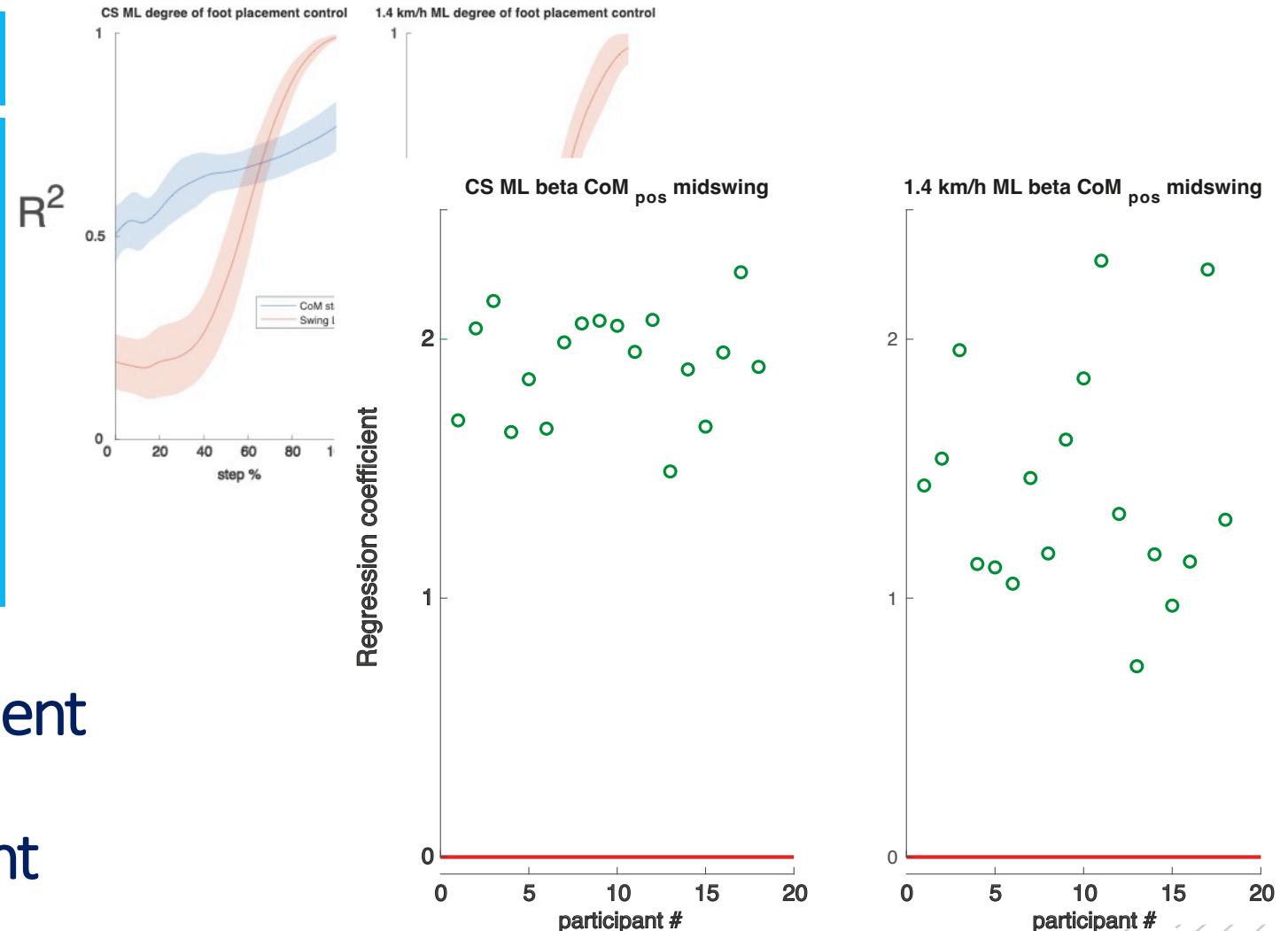
```
OUT.Combined_pct.data(i_pred_sample) = stats.rsquare;
```

```
[OUT,intermediates] =  
foot_placement_model_function_step(COM,Rfoot,Lfoot,events,fsopto,pred_samples,order,removeorigin,centerdata)
```

Foot placement model

Outcome measures

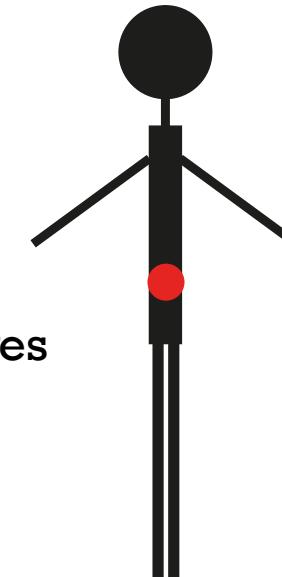
More lateral CoM displacement
leads to
more lateral foot placement



Foot placement model

Analysis pipeline

- Measuring kinematic data: Pelvis, heel markers
- Computing gait events
- Preparing the predictors (ML/AP CoM kinematic state)
- Preparing the dependent variable (step width/step length)
- Performing the linear regression
- Interpreting the outcome measures



Foot placement model

Key functions

- Gait events:
 - `calc_events_zeni`
 - `check_signal (VU3Dmodel)`
- Foot placement model
 - `foot_placement_model_function_step_ML`
 - `foot placement_model_fuction_step_AP`
 - `calc_derivative (VU3Dmodel)`
 - `normalizetimebase_step`
 - `regstats`

Foot placement model

Scripts

- **main1_gaitevents**
 - Compare two methods of gait event detection
 - Why would we use Zeni for Parkinson's patients?
- **main2_checkdata**
- **main3_FPmodel & main4_Fpplots**
 - Compare ML and AP results
 - Look at different outcome measures (R^2 , gains, foot placement error)
 - Look at the results for the predictor (CoM kinematic state) at different percentages of the swing phase.
 - At which percentage of the step cycle is the CoM kinematic state the most meaningful as a predictor?
 - Which outcome measure do you expect to improve for Parkinson's patients?

Final note

Coordinate system

- Stick to the consortium agreement
- Use same coordinate system for all subjects and trials
- Make a note of the coordinate system you define

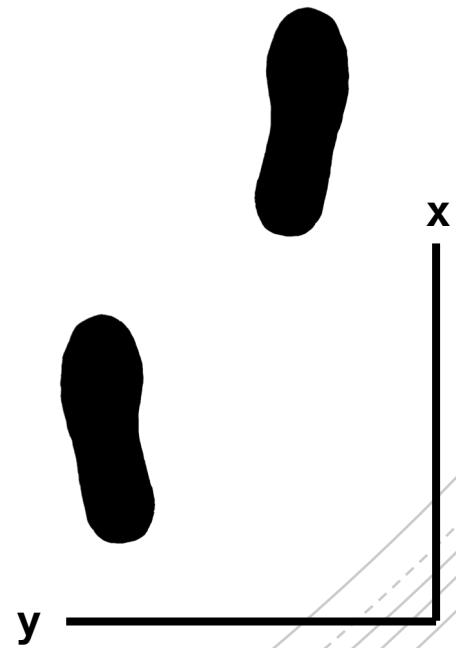
Today's dataset:

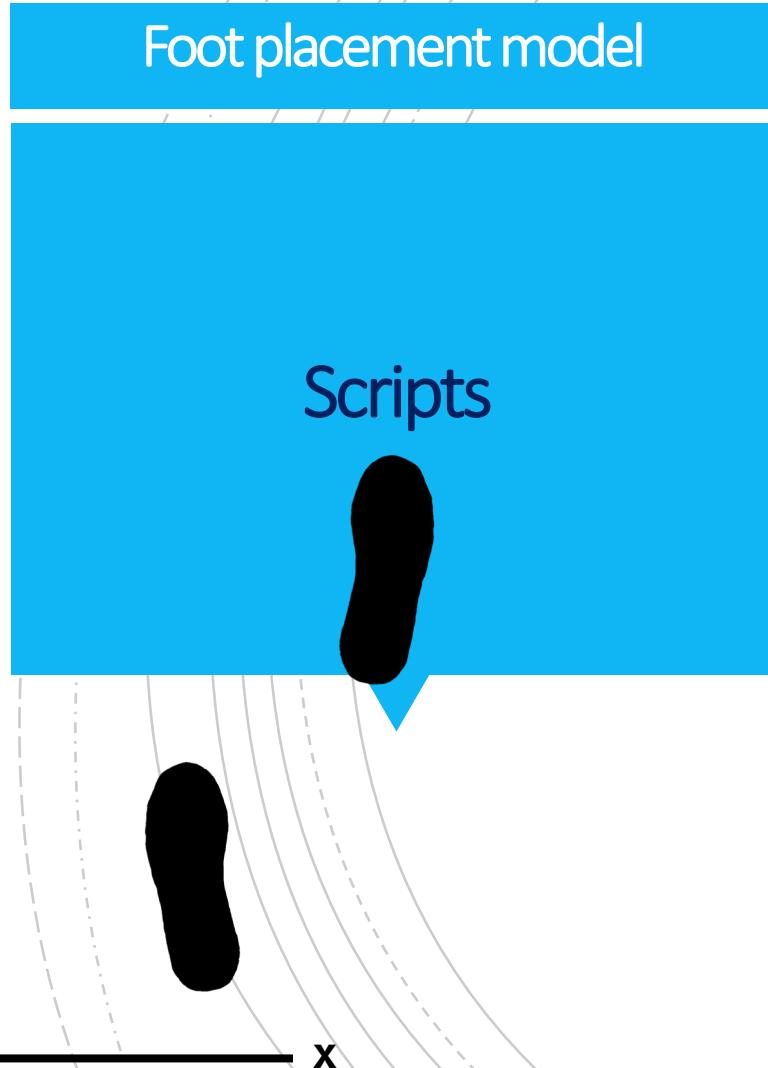
- x: mediolateral
(right +)
- y: anteroposterior
(forward +)
- z: vertical
(upward+)



Consortium agreement:

- x: anteroposterior
(forward +)
- y: mediolateral
(left +)
- z: vertical
(upward+)





- **main1_gaitevents**
 - Compare two methods of gait event detection
 - Why would we use Zeni for Parkinson's patients?
- **main2_checkdata**
- **main3_FPmodel & main4_Fpplots**
 - Compare ML and AP results
 - Look at different outcome measures (R^2 , gains, foot placement error)
 - Look at the results for the predictor (CoM kinematic state) at different percentages of the swing phase.
 - At which percentage of the step cycle is the CoM kinematic state the most meaningful as a predictor?
 - Which outcome measure do you expect to improve for Parkinson's patients?

Amsterdam team: prof. dr. Jaap van Dieën, dr. Sjoerd Bruijn, dr. Bernadette van Wijk
dr. Moira van Leeuwen

Foot Placement Control & Corticomuscular Coherence

Concepts &
Analysis Pipeline

Amsterdam
Movement
Sciences

VU VRIJE
UNIVERSITEIT
AMSTERDAM

 Amsterdam UMC
University Medical Centers

Amsterdam team: prof. dr. Jaap van Dieën, dr. Sjoerd Bruijn, dr. Bernadette van Wijk
dr. Moira van Leeuwen

Foot Placement Control & Corticomuscular Coherence

Concepts &
Analysis Pipeline

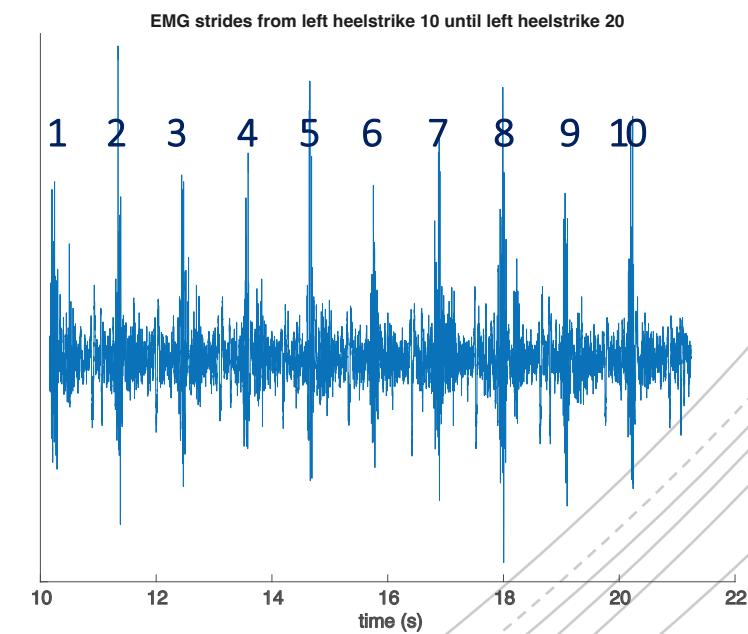
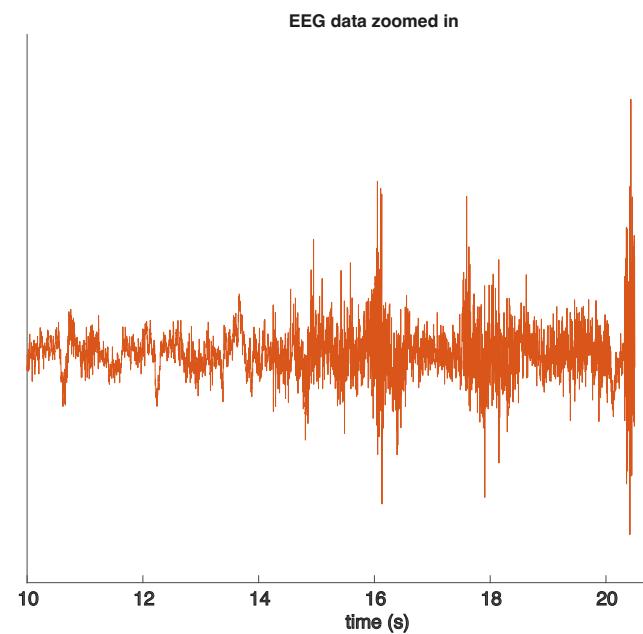
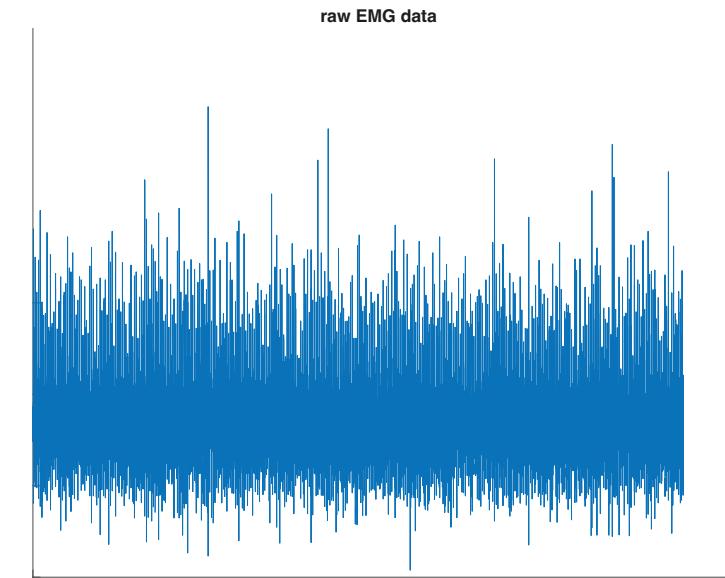
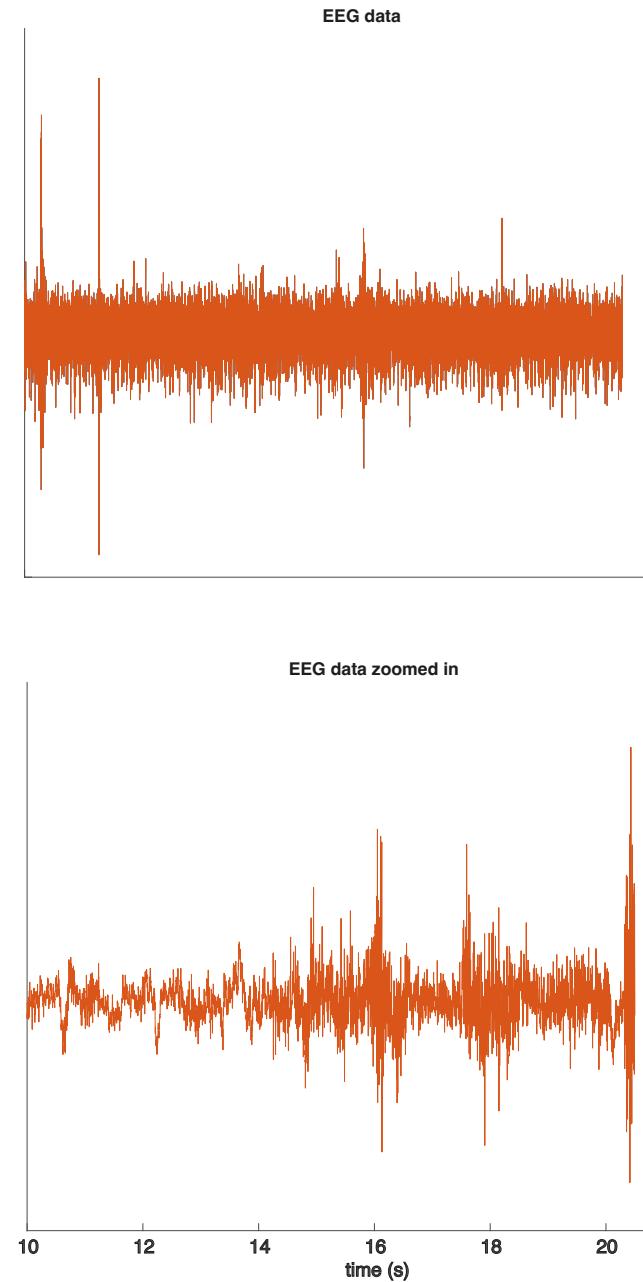
Amsterdam
Movement
Sciences

VU VRIJE
UNIVERSITEIT
AMSTERDAM

U Amsterdam UMC
University Medical Centers

Corticomuscular coherence

EMG
&
EEG data



plot_stride_av_EMG_ft

```
muscle_name =    
norm_val = 100;
```

```
plot_stride_av_EMG_ft(muscle_name,emg_data,events_EMG,norm_val)
```

Corticomuscular coherence

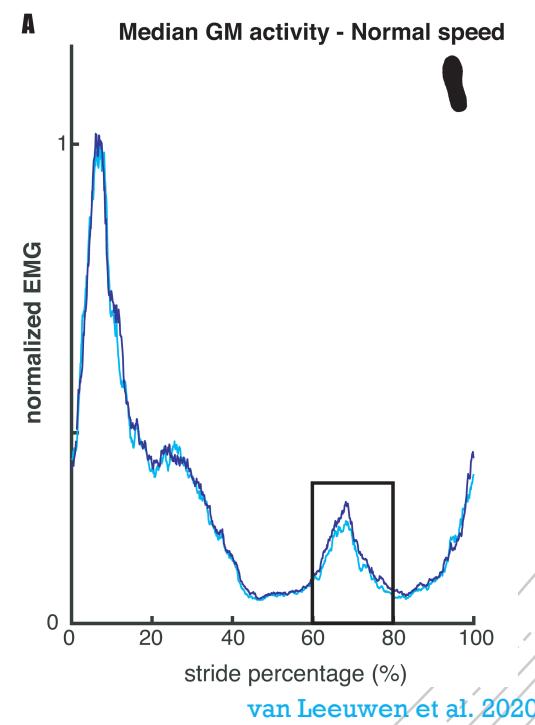
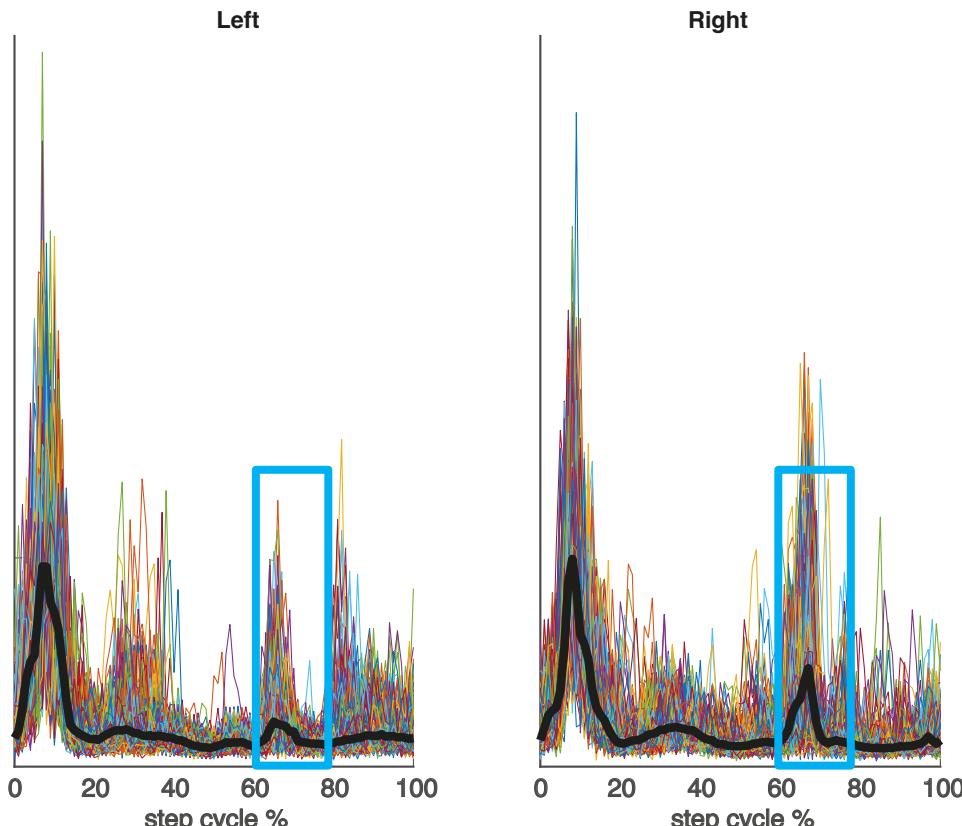
Quick check: EMG

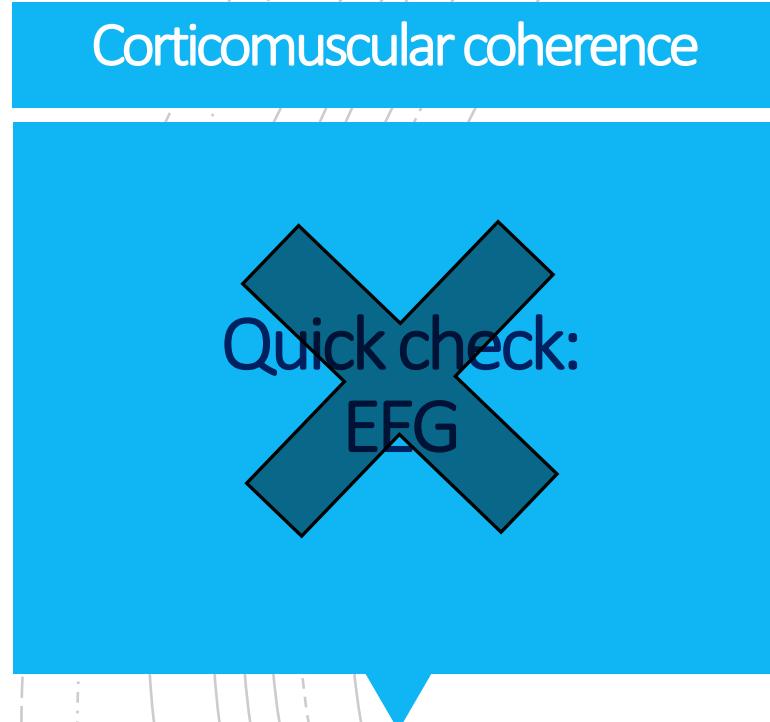
```
[EMG] =  
EMG_preprocessing(EMG_R,events_EMG.fs,20,50); %  
conform Rankin et al. (2014)
```

High pass filter → Rectification → Low pass filter

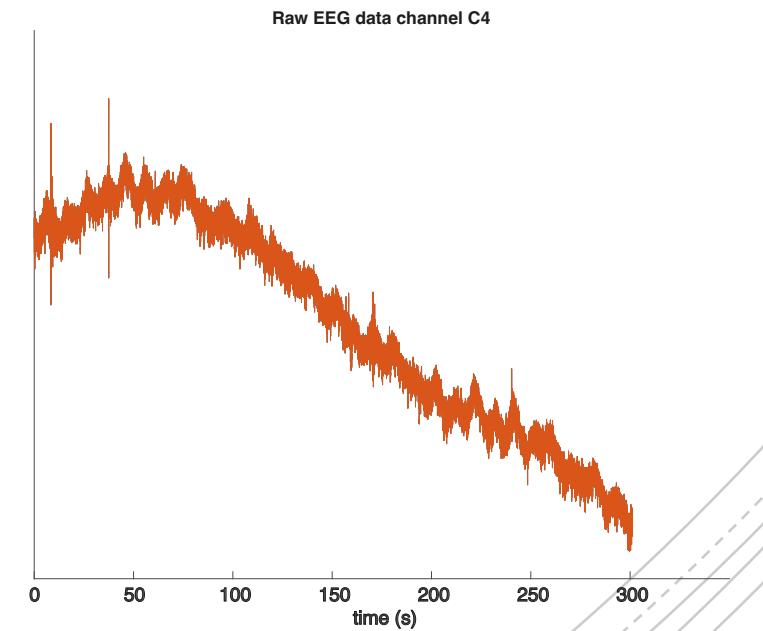
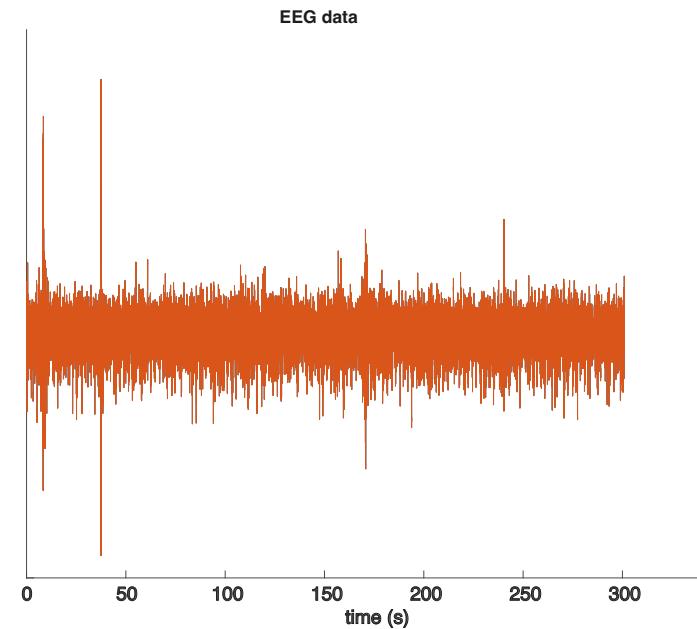
normalize_gait_events

Time normalization from heelstrike to ipsilateral heelstrike





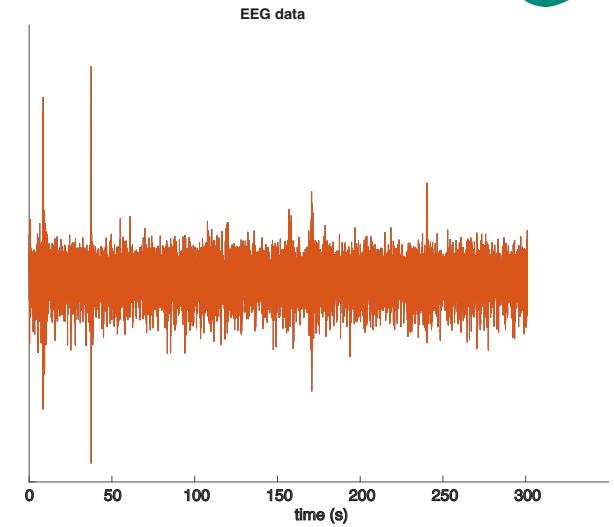
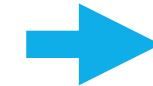
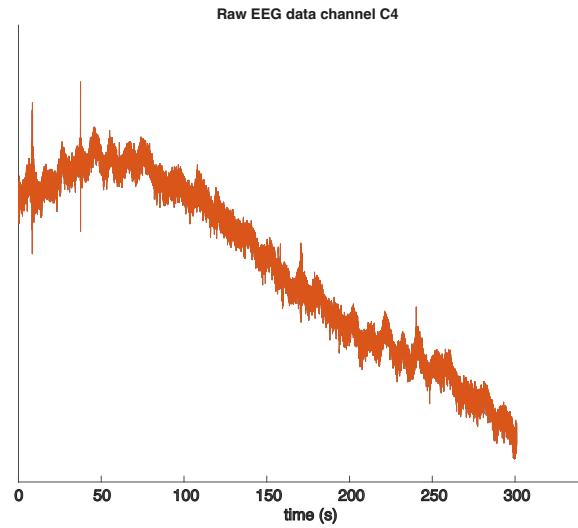
- Preprocessing
- Identifying artefacts ICA



ft_preprocessing

Corticomuscular coherence

EEG
preprocessing



% Filters

```
cfg.bpfILTER= 'yes'  
cfg.bpfILTORD = 2;
```

```
cfg.bpfreq = [0.5 220];
```

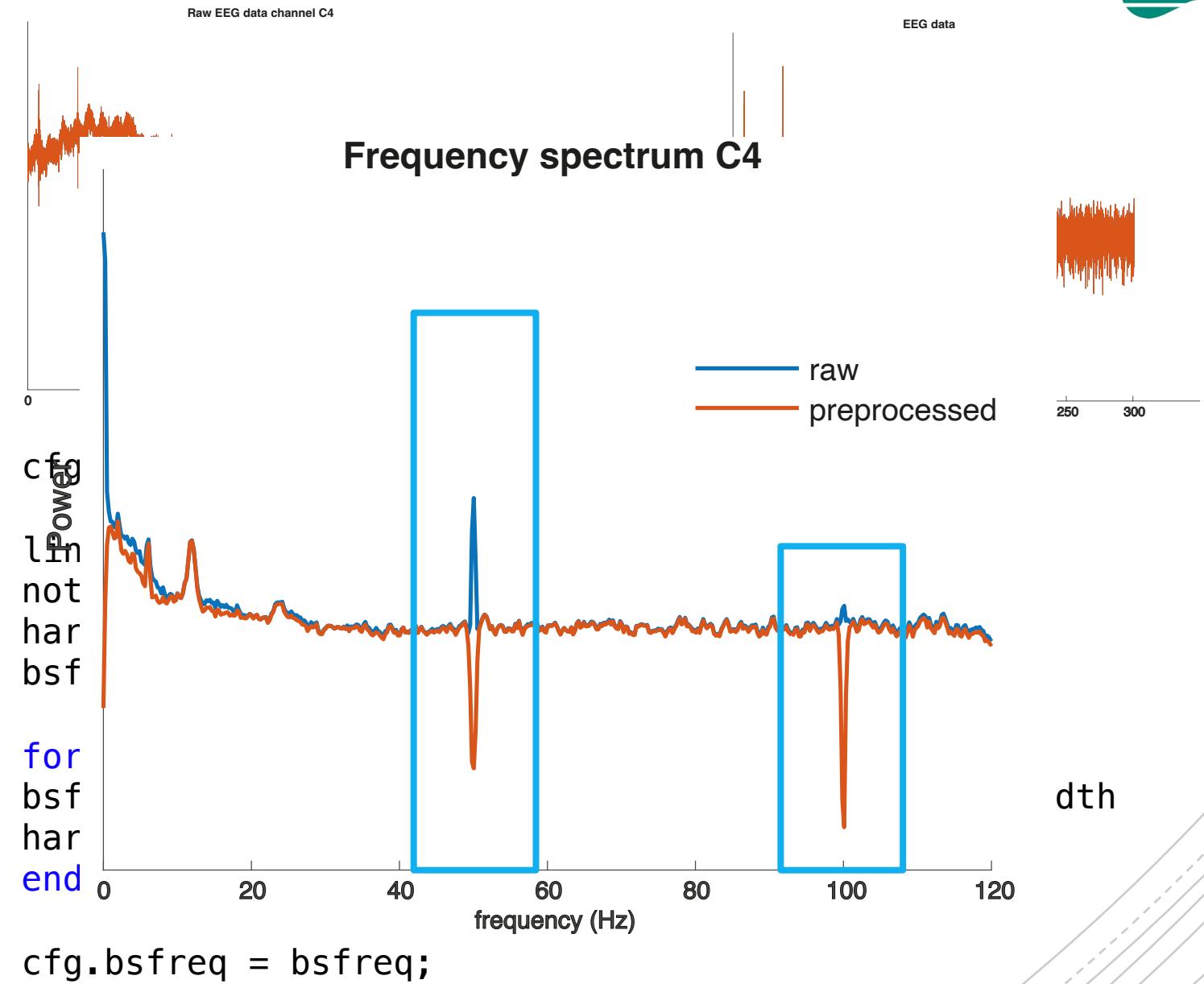
% Demean & detrend

```
cfg.demean = 'yes';  
cfg.detrend = 'yes';
```

ft_preprocessing

Corticomuscular coherence

EEG
preprocessing



Corticomuscular coherence

EEG Bad channels

- “Bad channels with a SD of their activation exceeding the mean SD activation plus 3 standard deviations” (Klapprott & Debener 2024)
- Single-channel variance exceeding 3 times the standard deviation, or 10 times the maximum of all channels) (Zandvoort 2019)
- Repairing bad channels based on interpolation neighboring electrodes

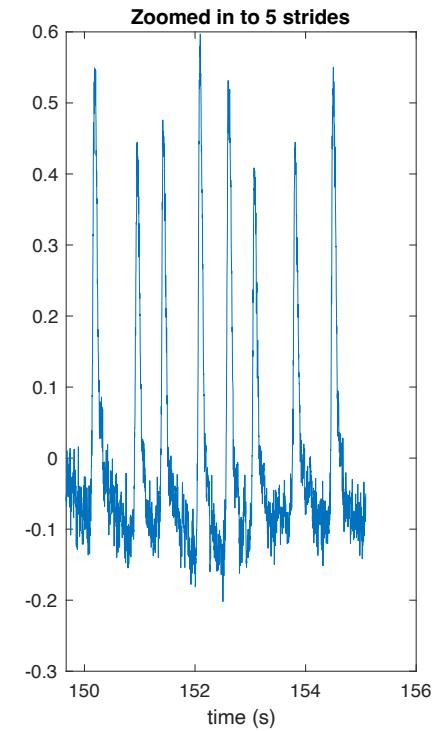
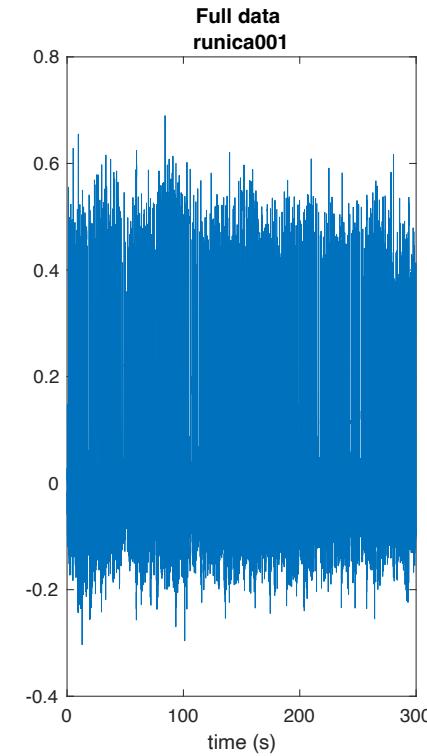
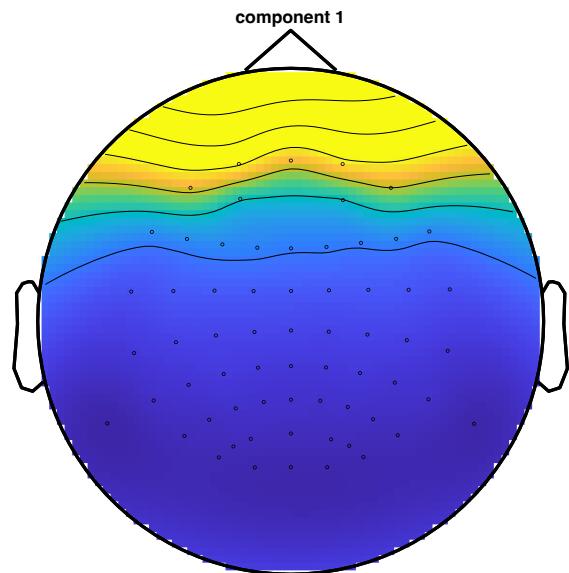
Corticomuscular coherence

ICA artefact removal

- Eye movement artefacts
- Muscle activity artefacts
- (line noise artefacts)

- EEG components

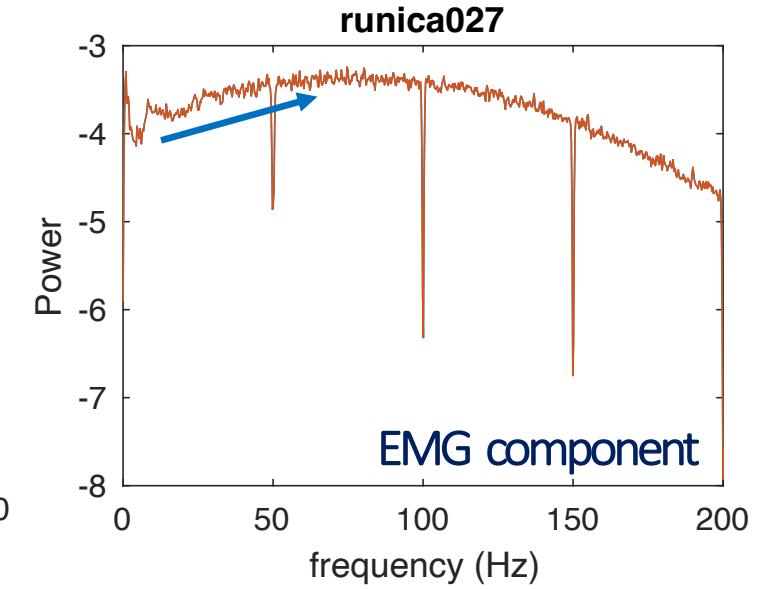
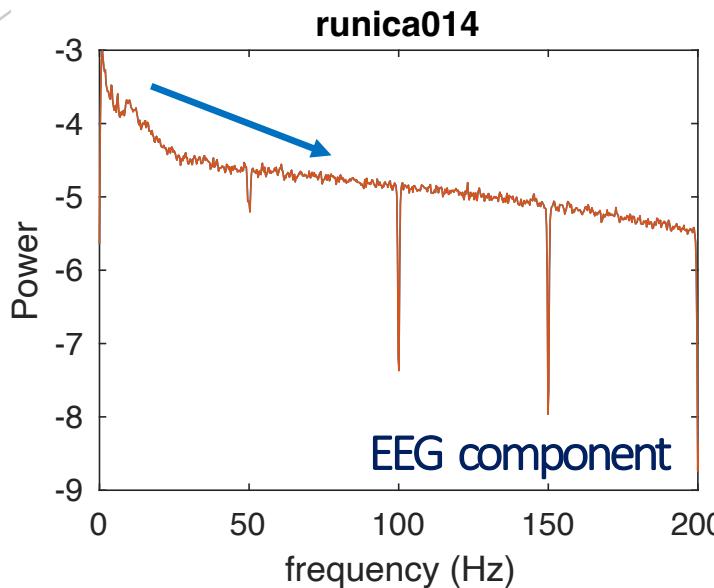
```
artifactChannels={ 'Fp1', 'Fp2', 'Fpz', 'AF7', 'AF3', 'AF4', 'AF8' };  
%frontal
```



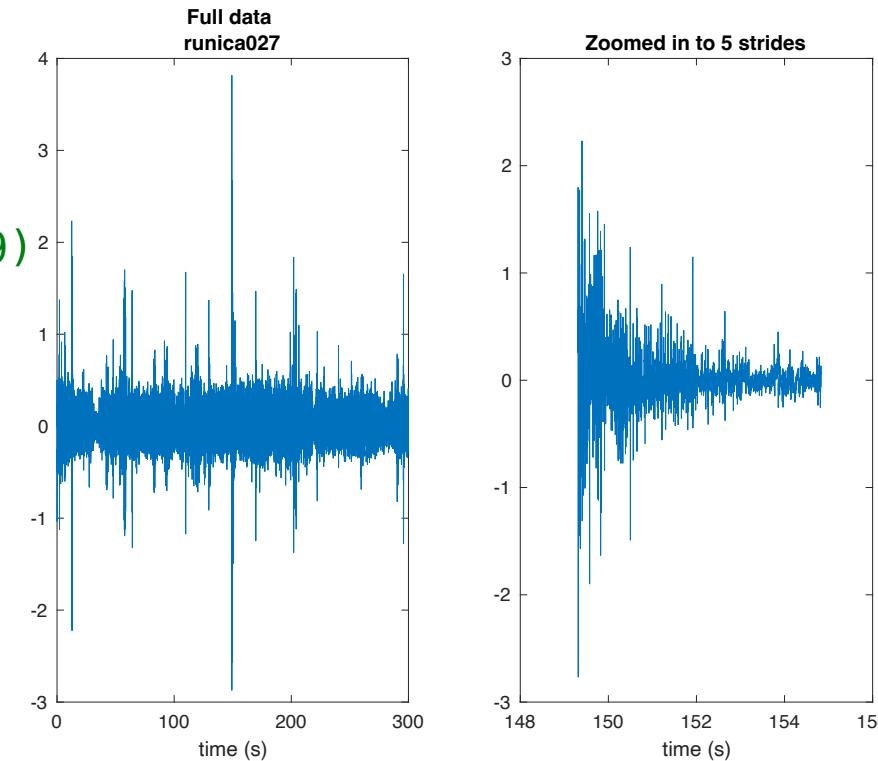
```
eye_artefact_comp1 = find(median_freq<5);
```

Corticomuscular coherence

ICA
Muscle activity
artefacts

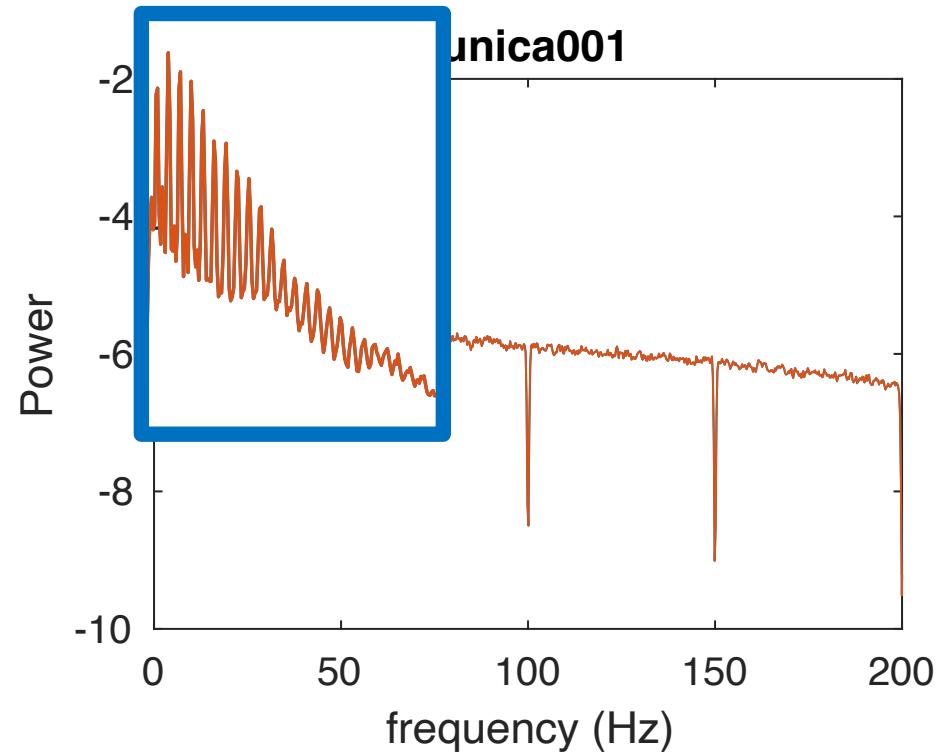


% Zandvoort et al (2019)
muscle_artefact_comp =
find(median_freq>60);

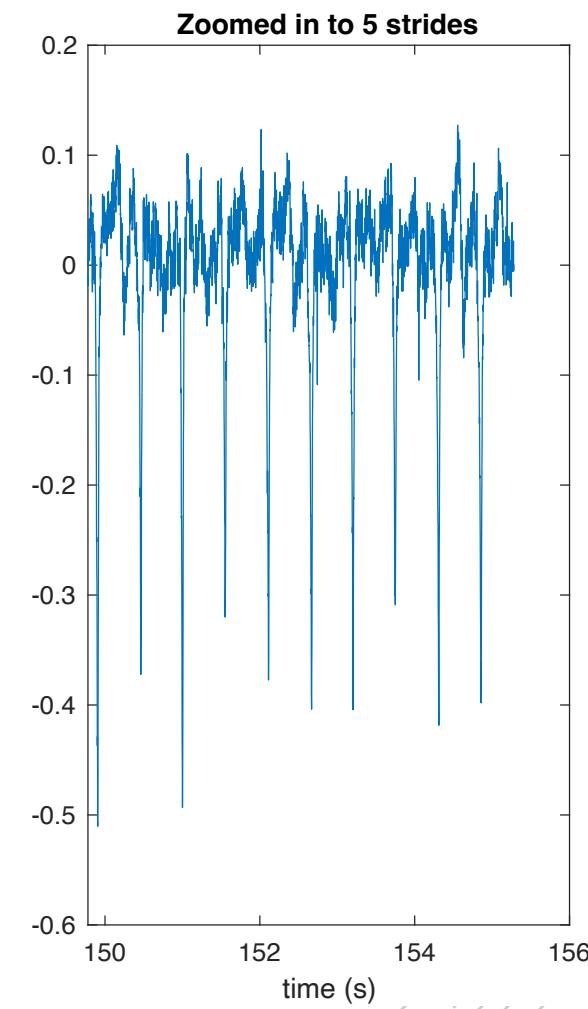


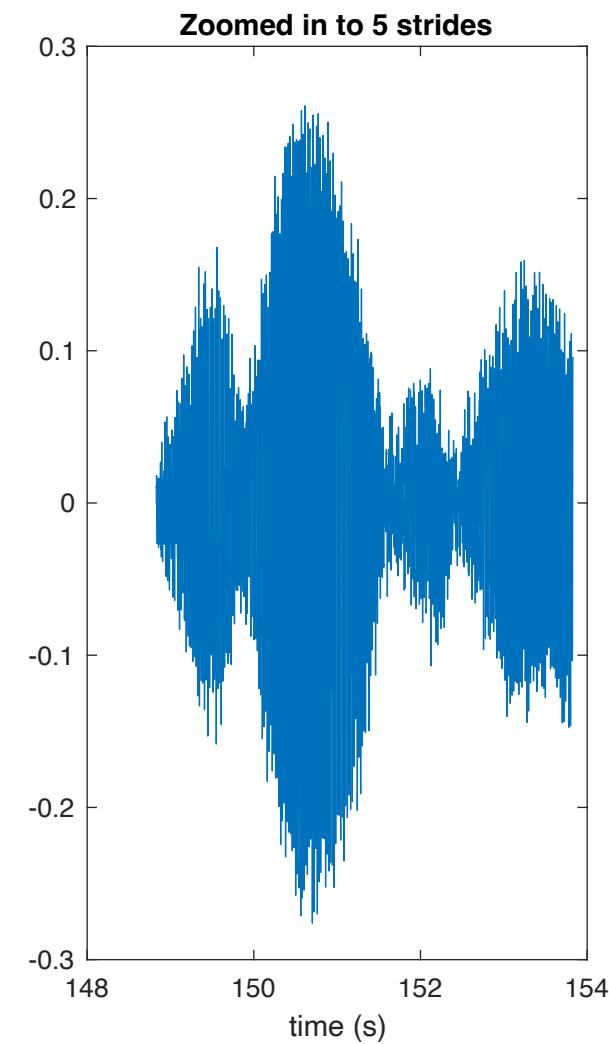
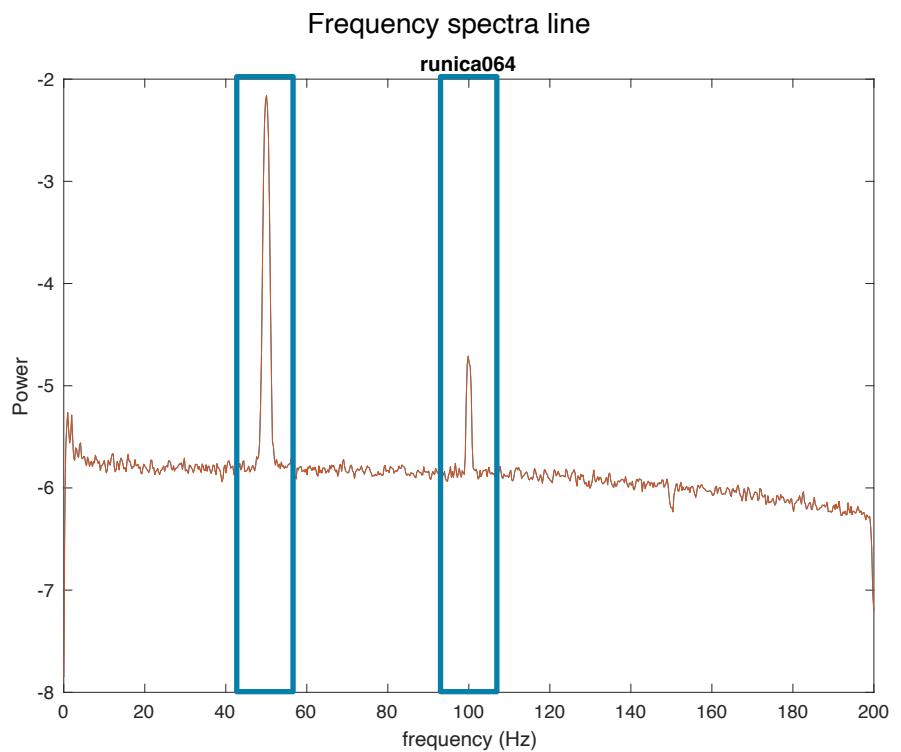
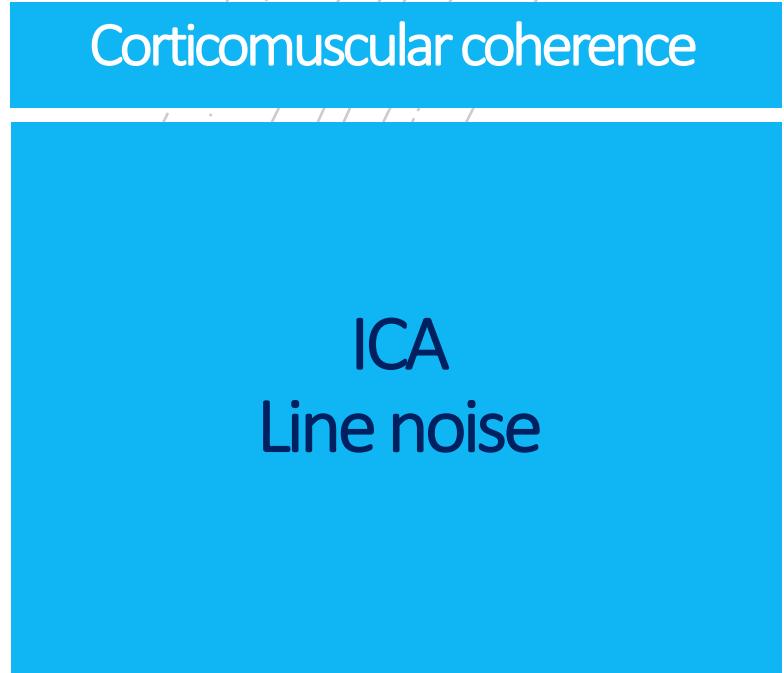
Corticomuscular coherence

ICA
Movement
artefacts



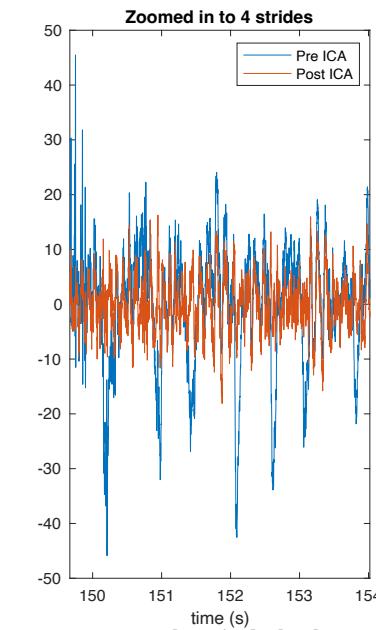
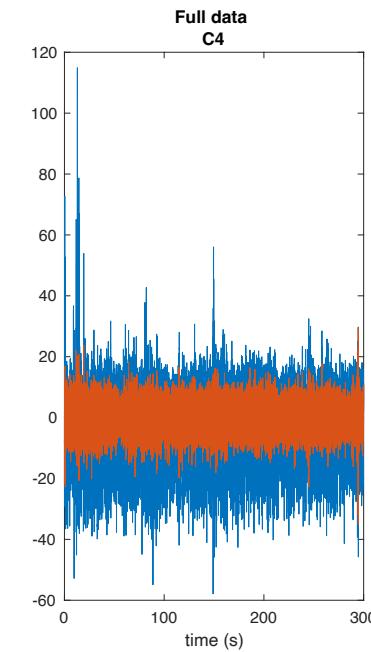
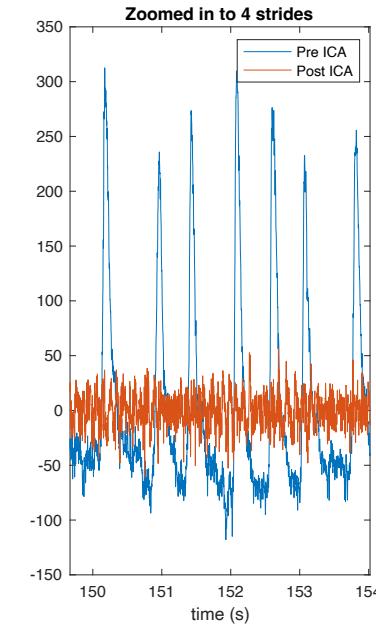
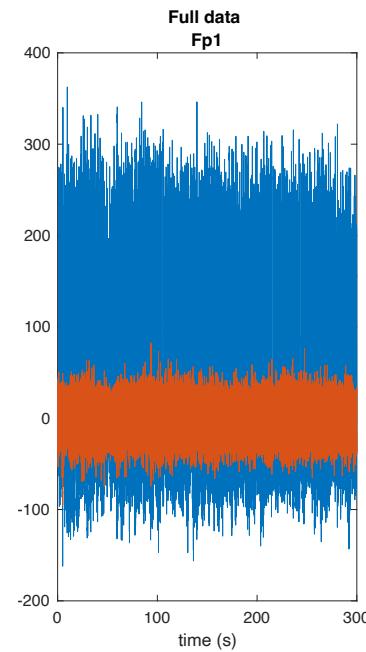
Stride frequency and
harmonics





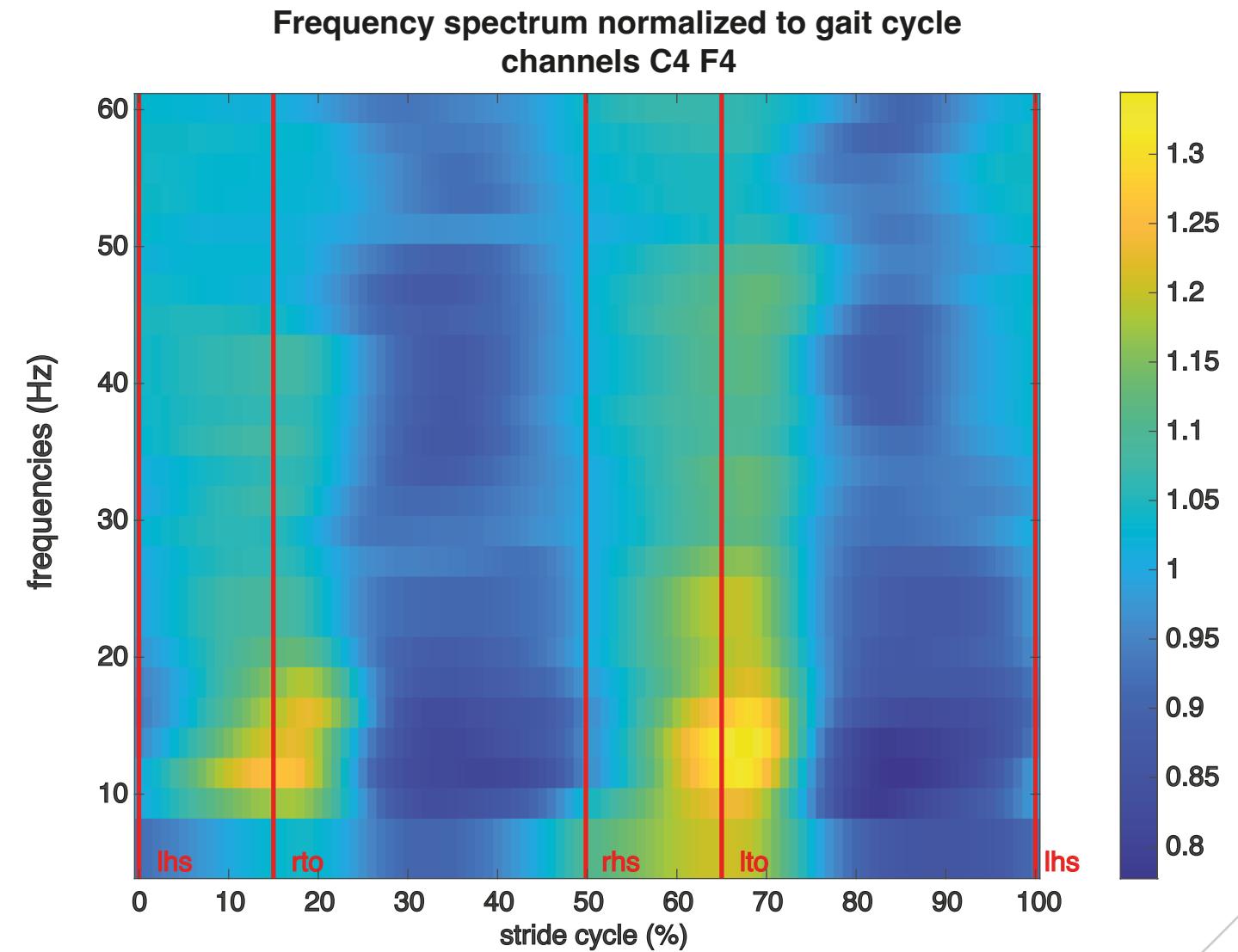
ICA effect on EEG

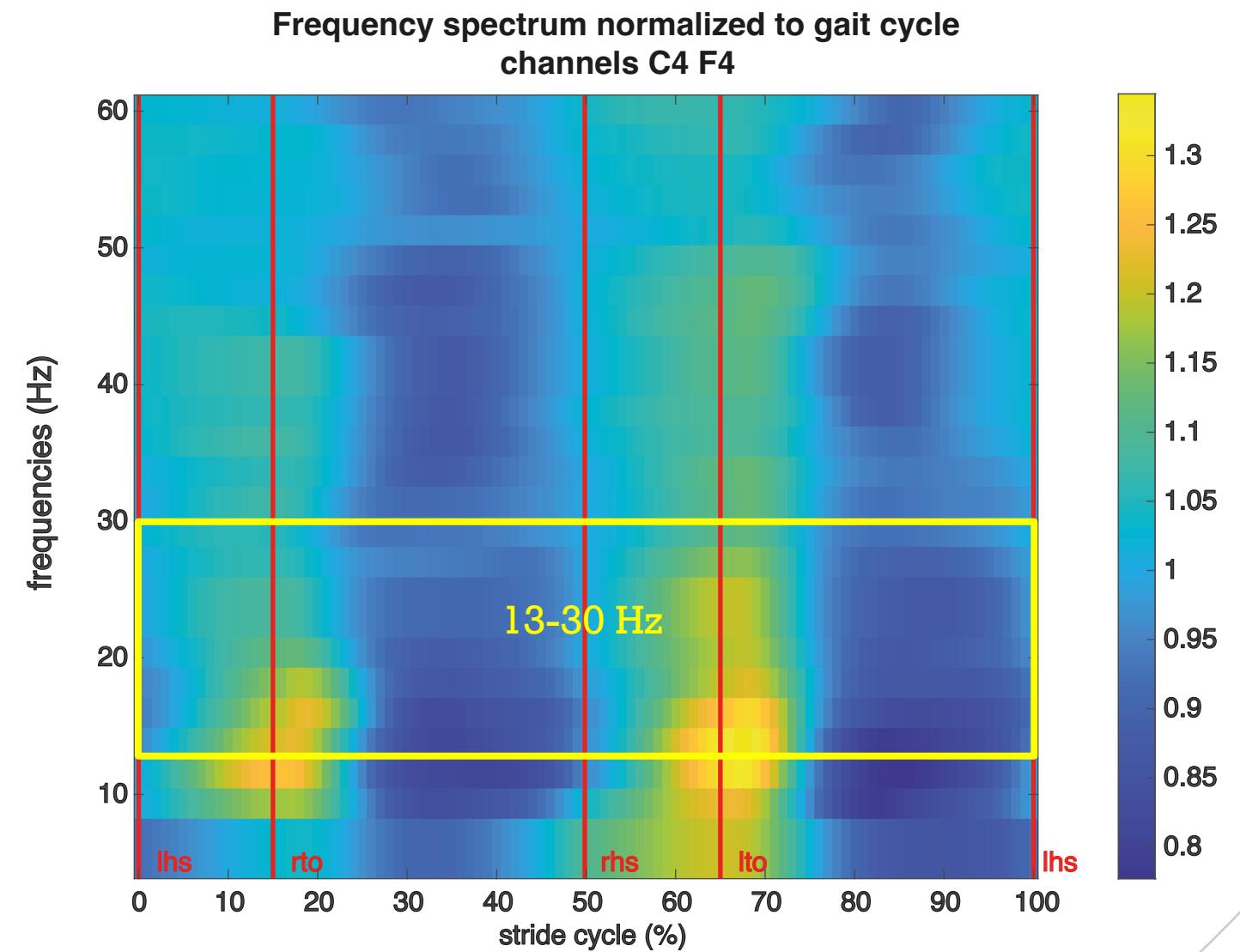
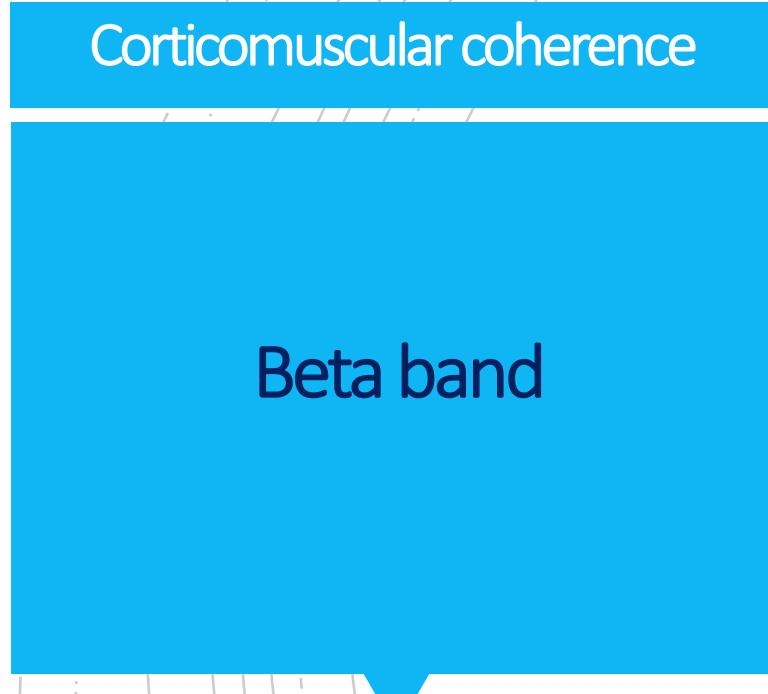
Corticomuscular coherence



ft_freqanalysis

Corticomuscular coherence





Corticomuscular coherence

Beta band

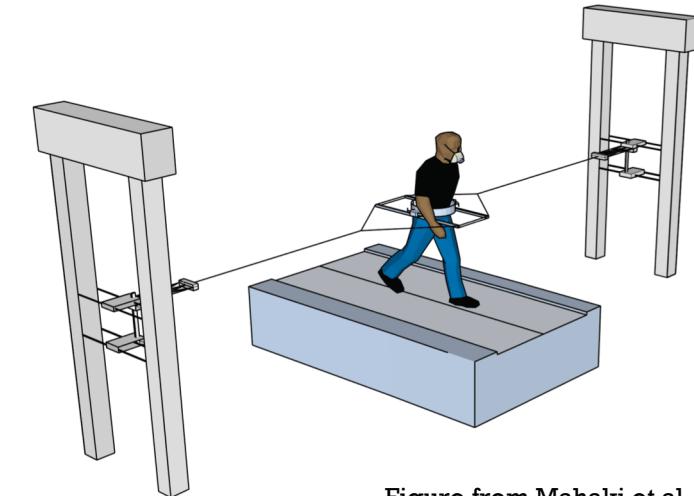
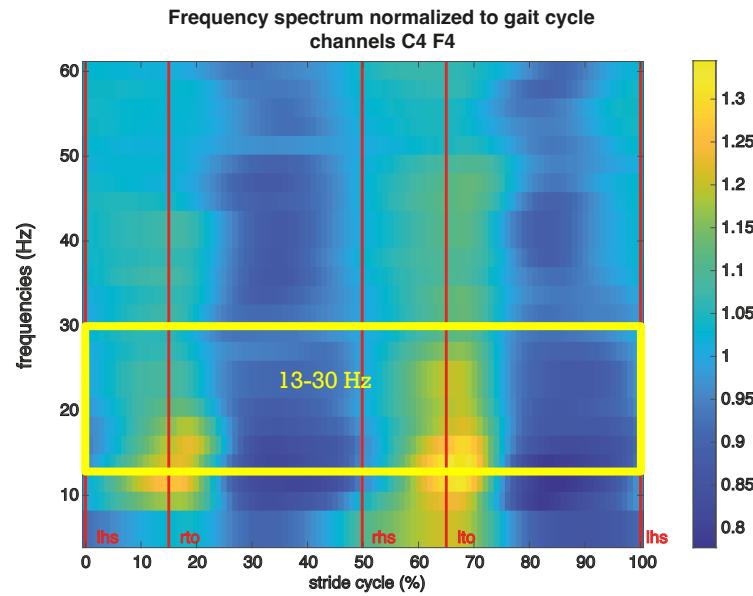
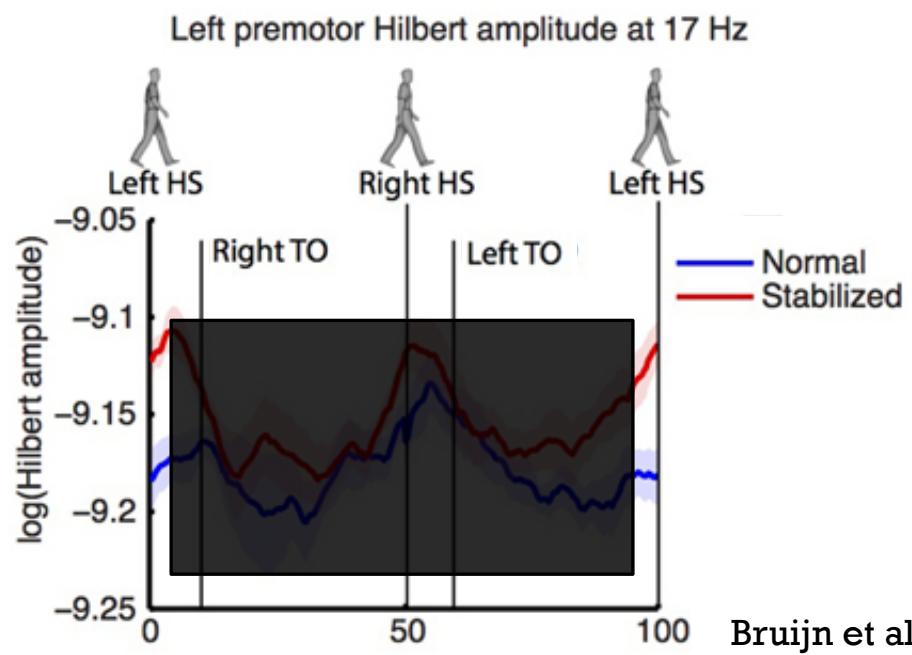
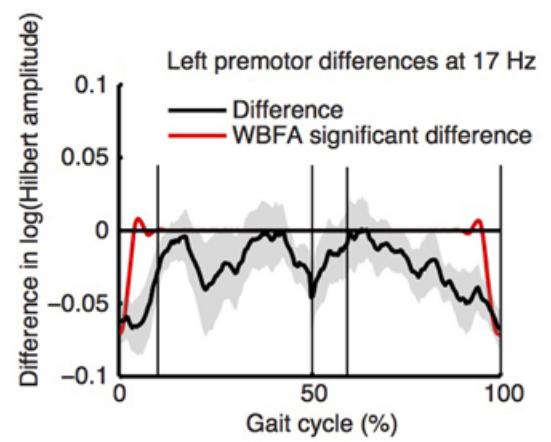


Figure from Mahaki et al.



Bruijn et al. 2015

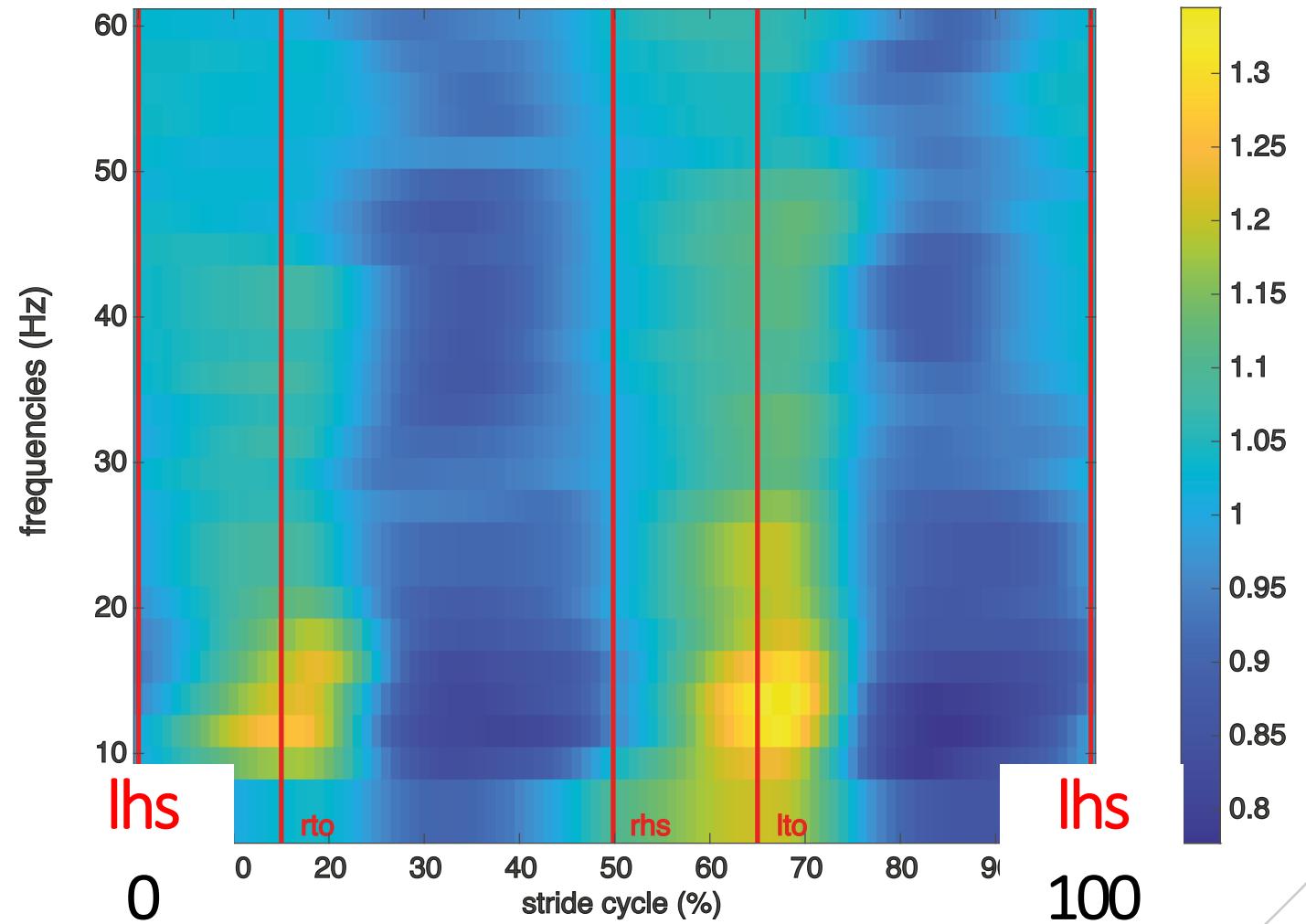


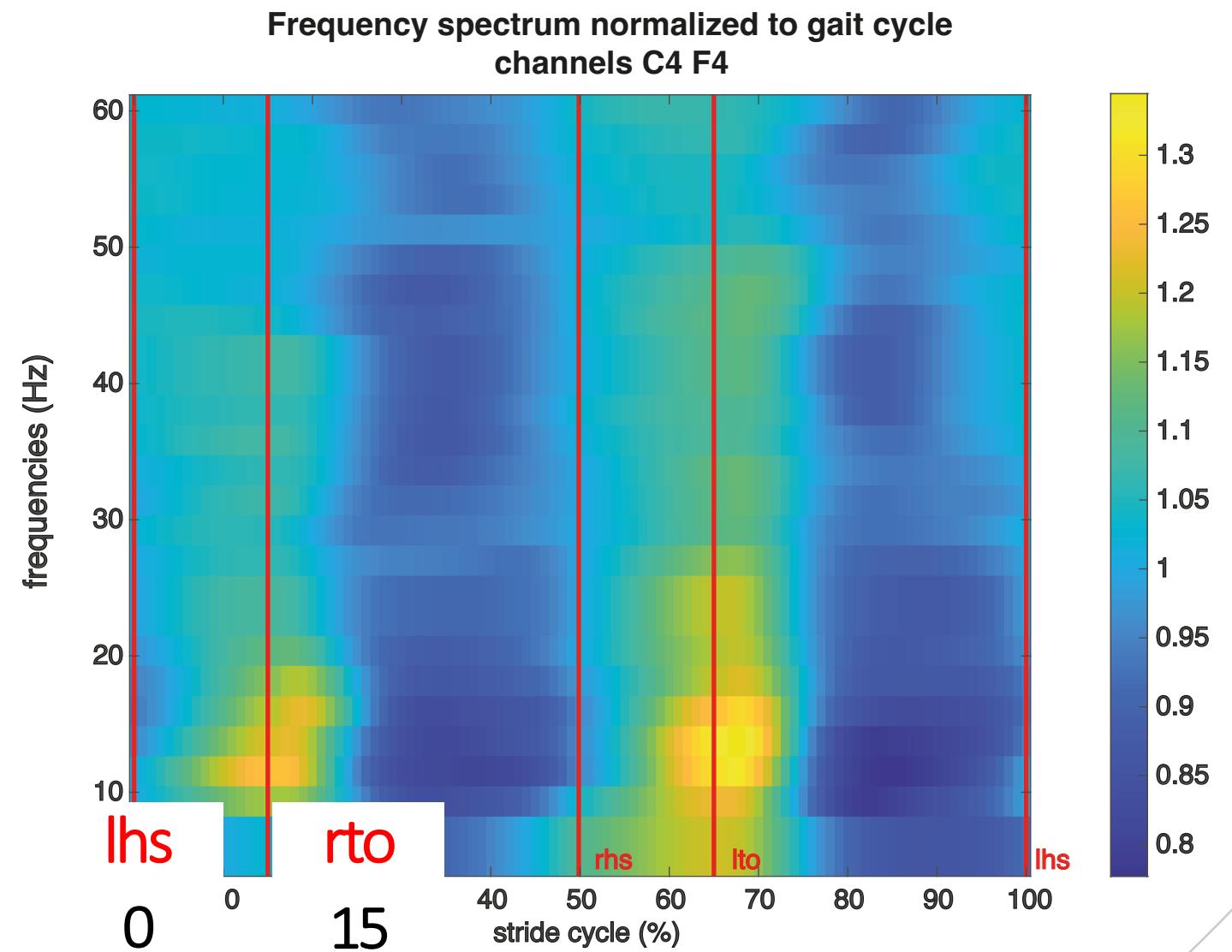
time_normalize_power

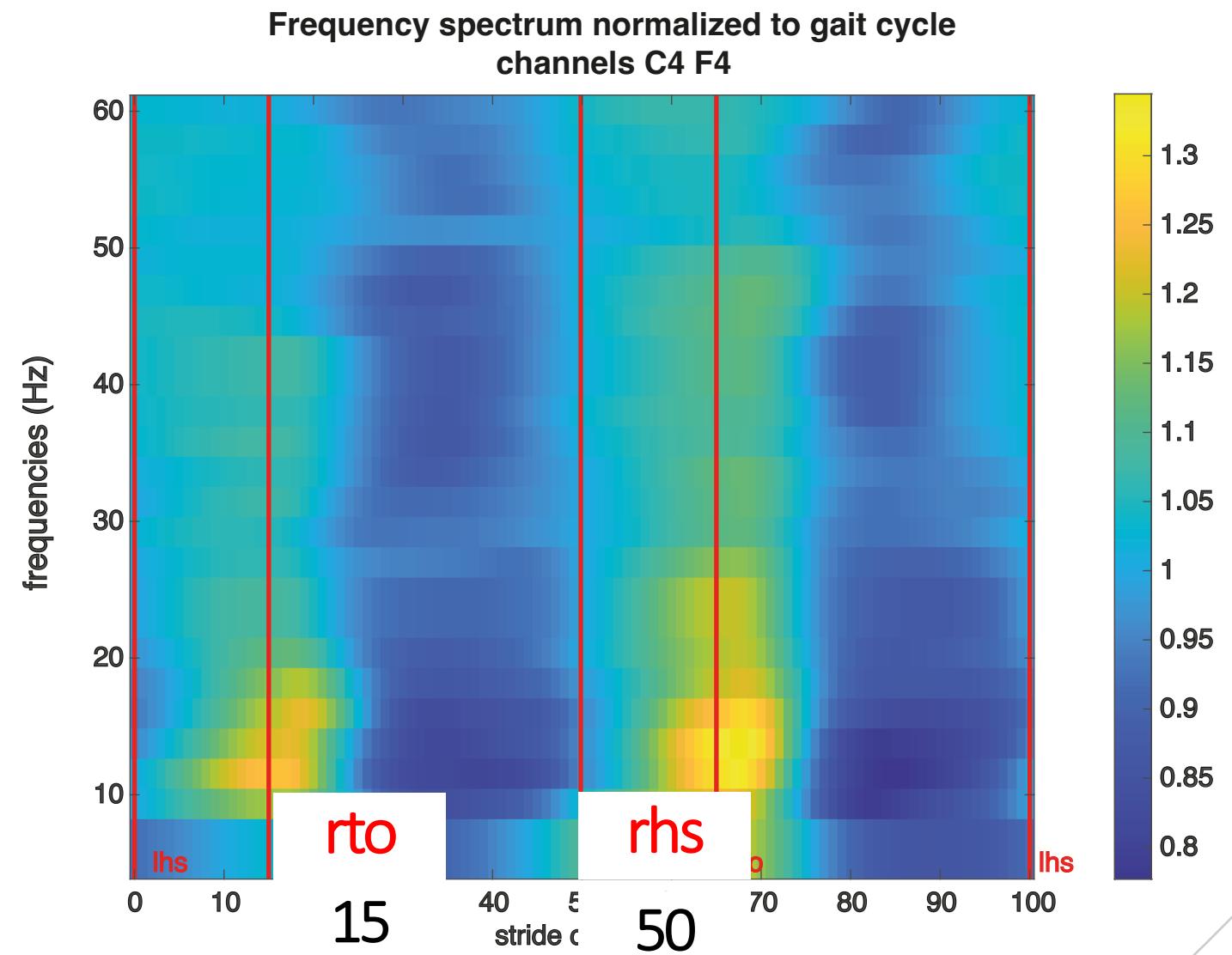
Corticomuscular coherence

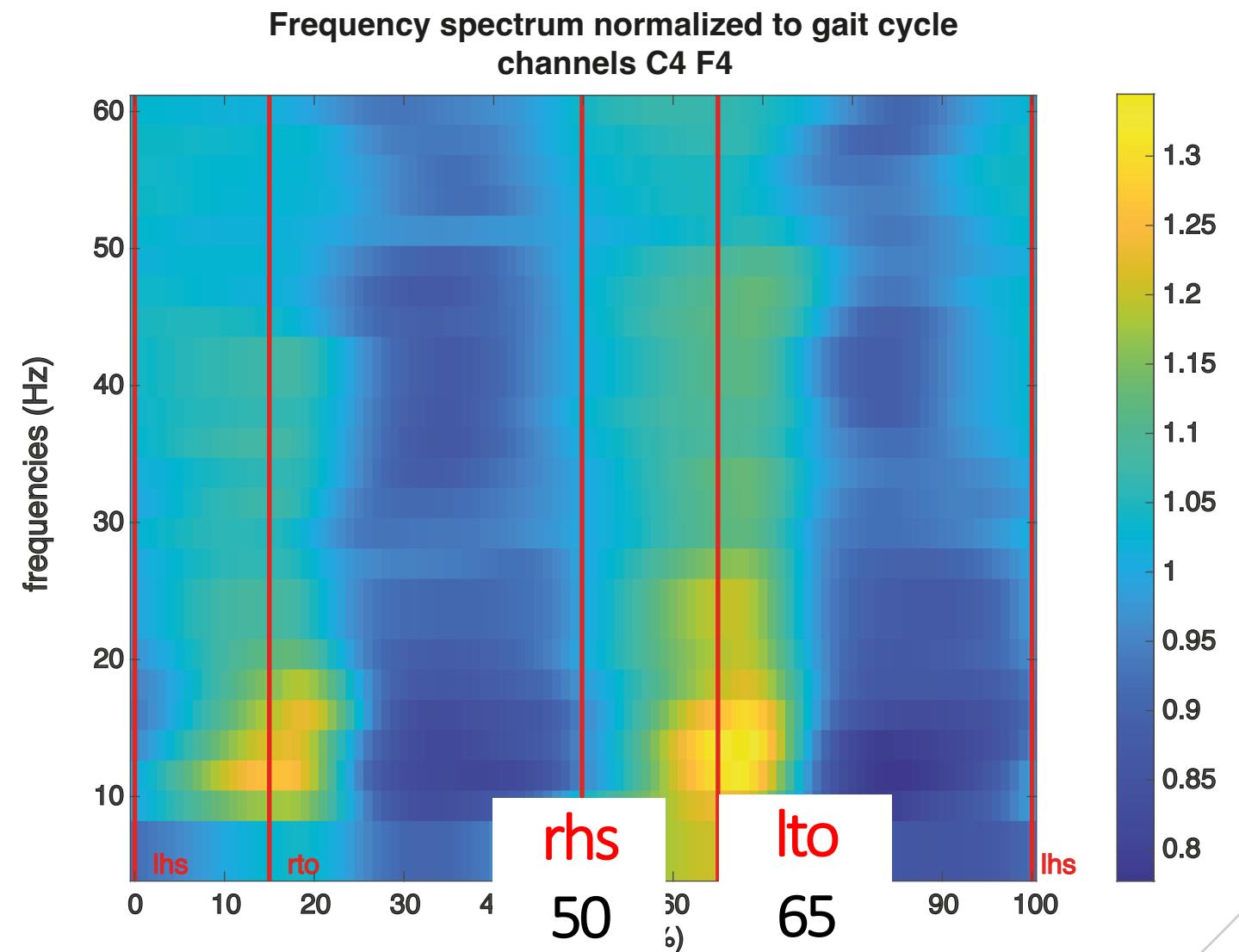
Power spectrum

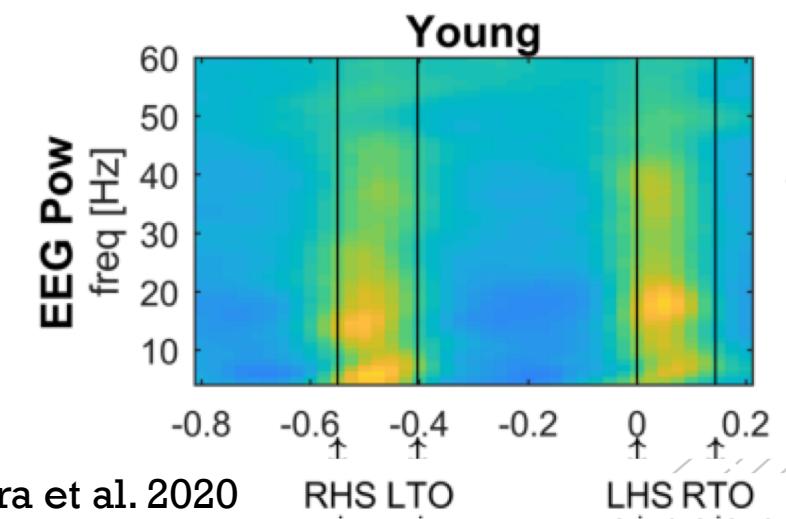
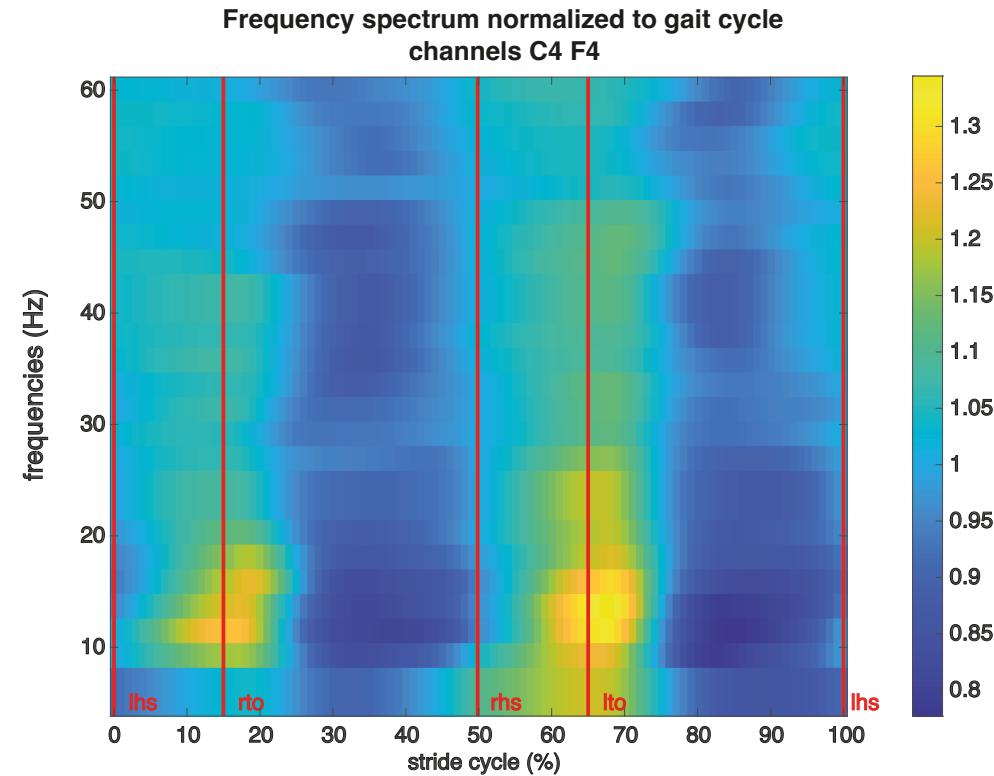
Frequency spectrum normalized to gait cycle
channels C4 F4



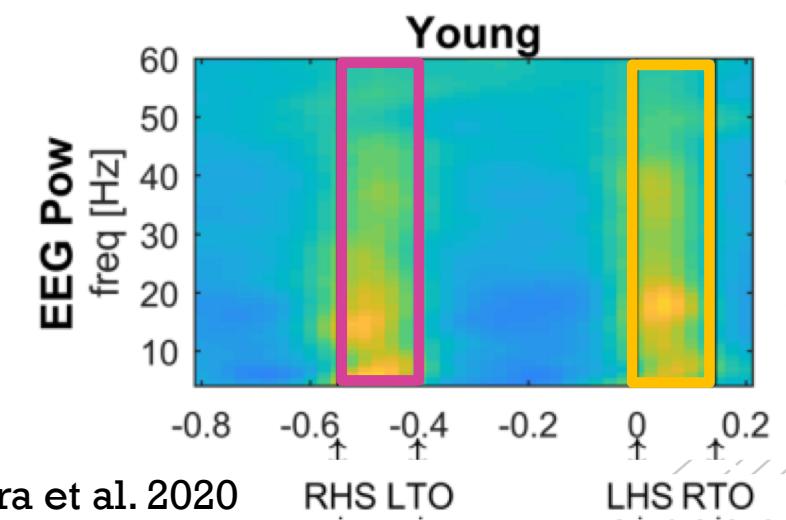
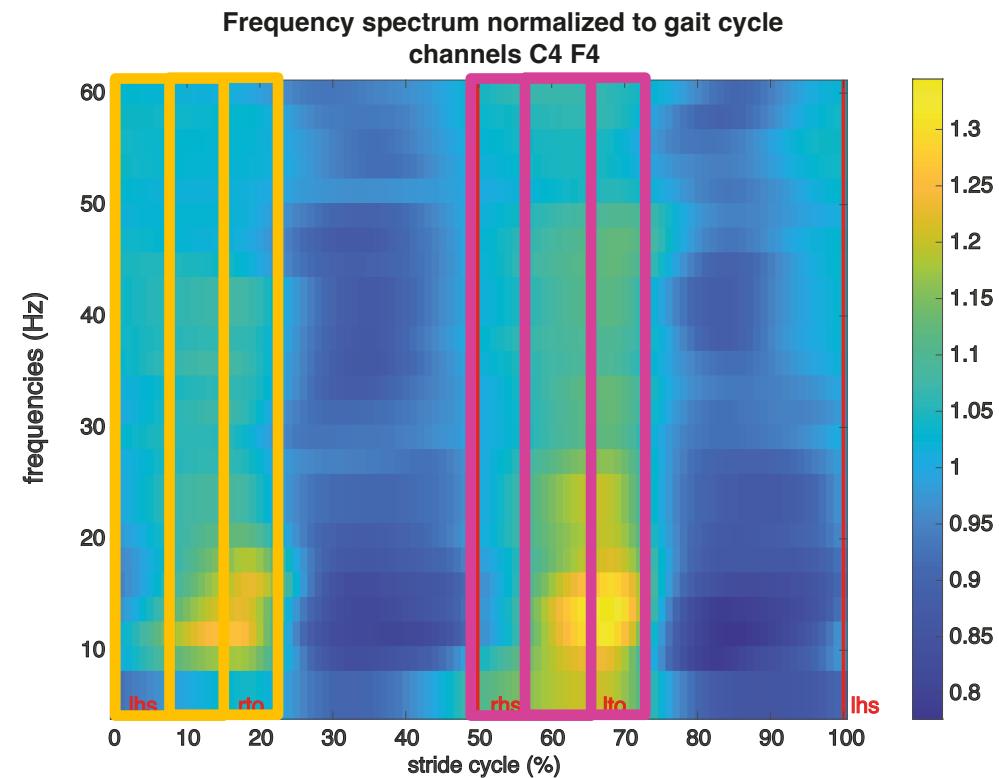






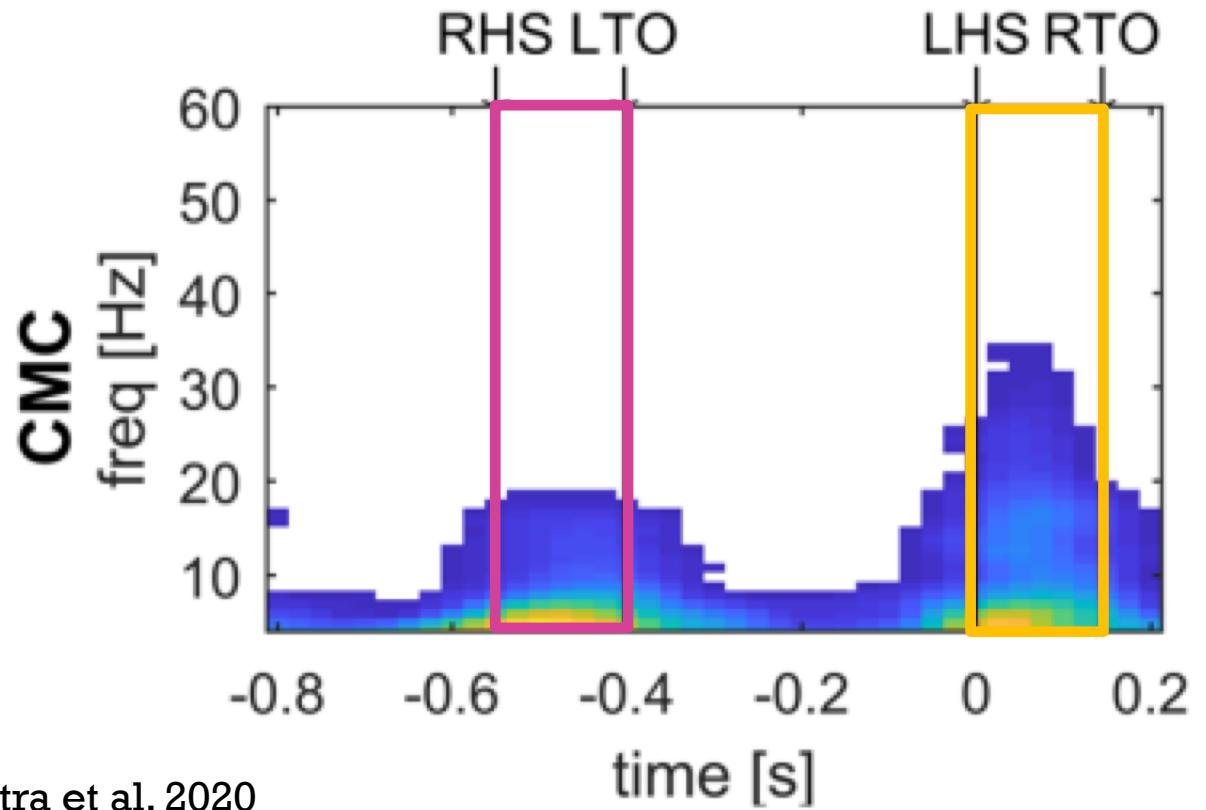
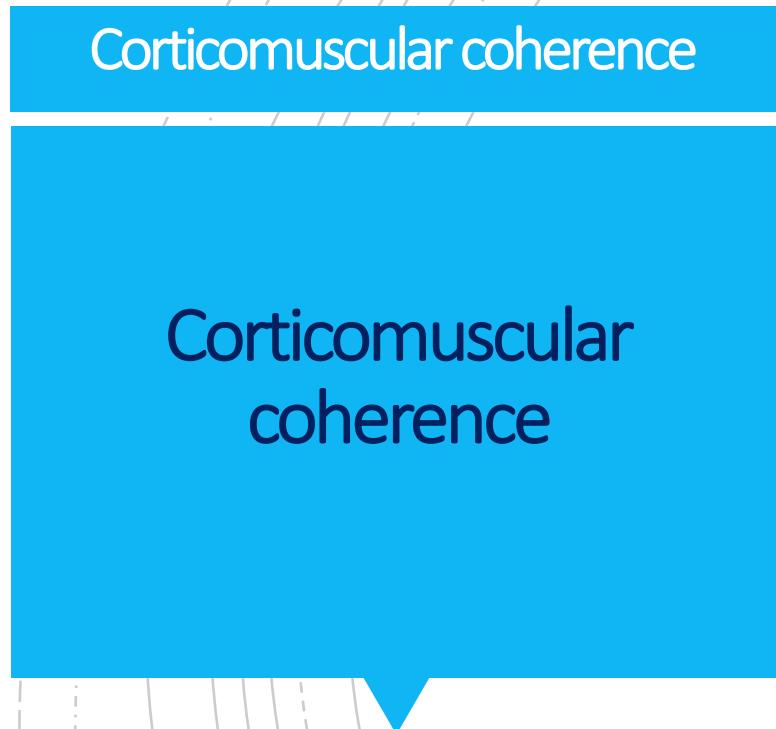


Boonstra et al. 2020



Boonstra et al. 2020

Tibialis Anterior C4- F4



Boonstra et al. 2020

Corticomuscular coherence

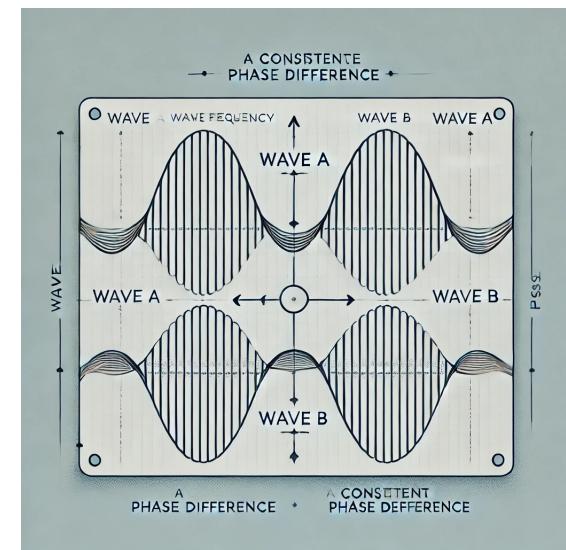
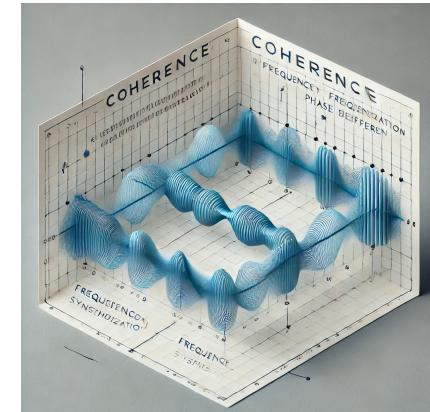
Corticomuscular
coherence

$$Coh_{S1,S2}(f) = \frac{|P_{S1,S2}(f)|^2}{|P_{S1}(f)| \times |P_{S2}(f)|}$$

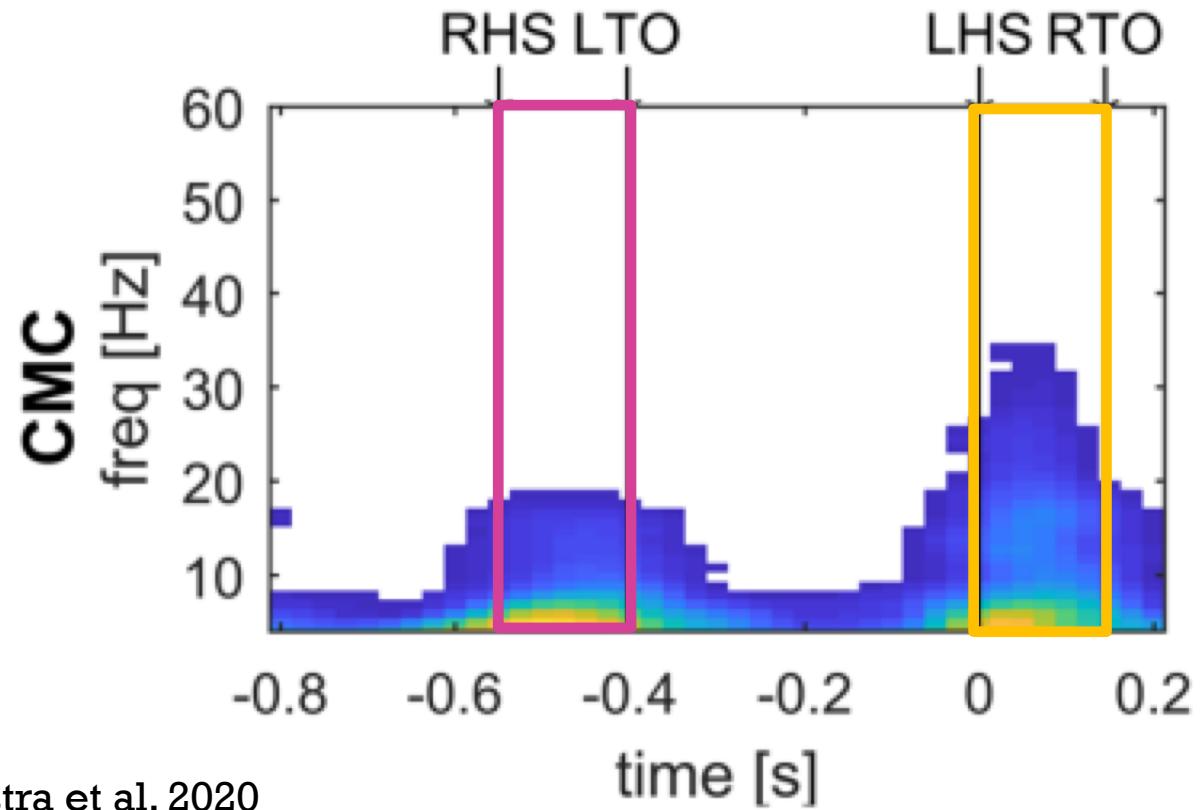
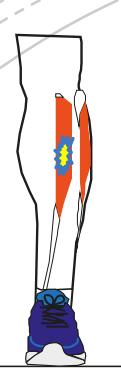
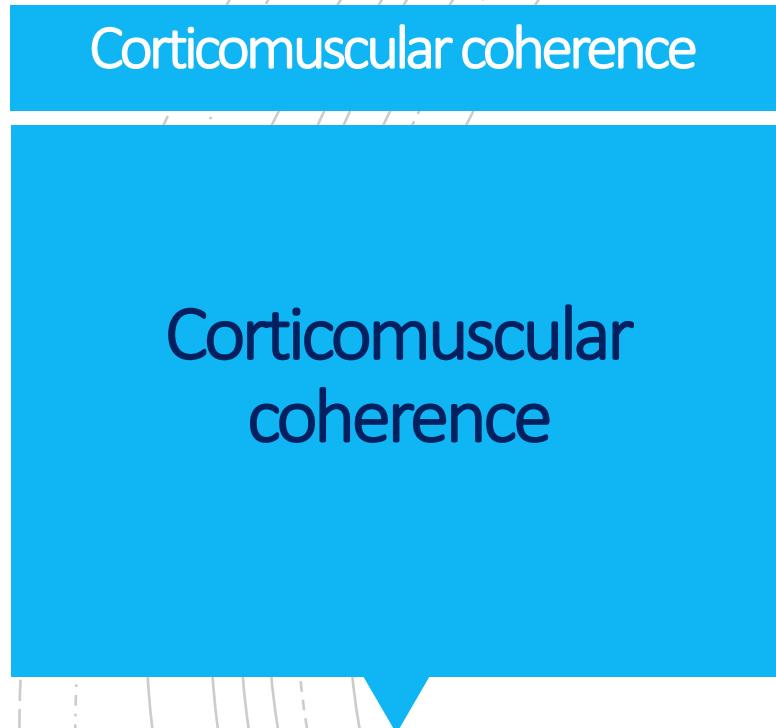
Functional connectivity

Cortical control over
muscle activity

Higher coherence
More/better stability control?

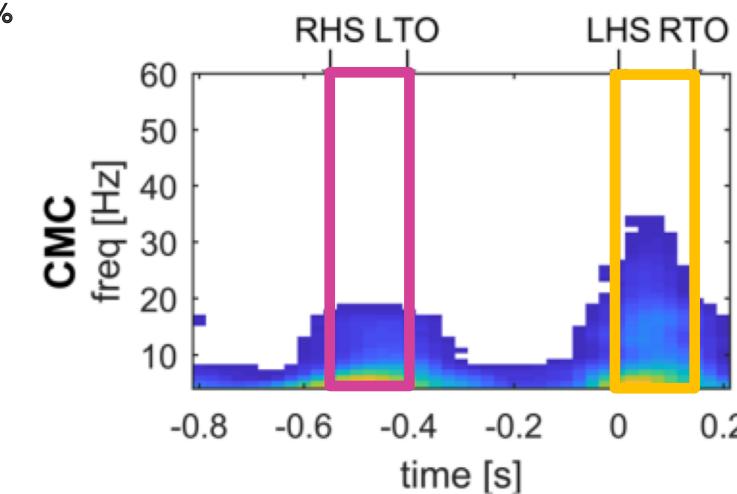
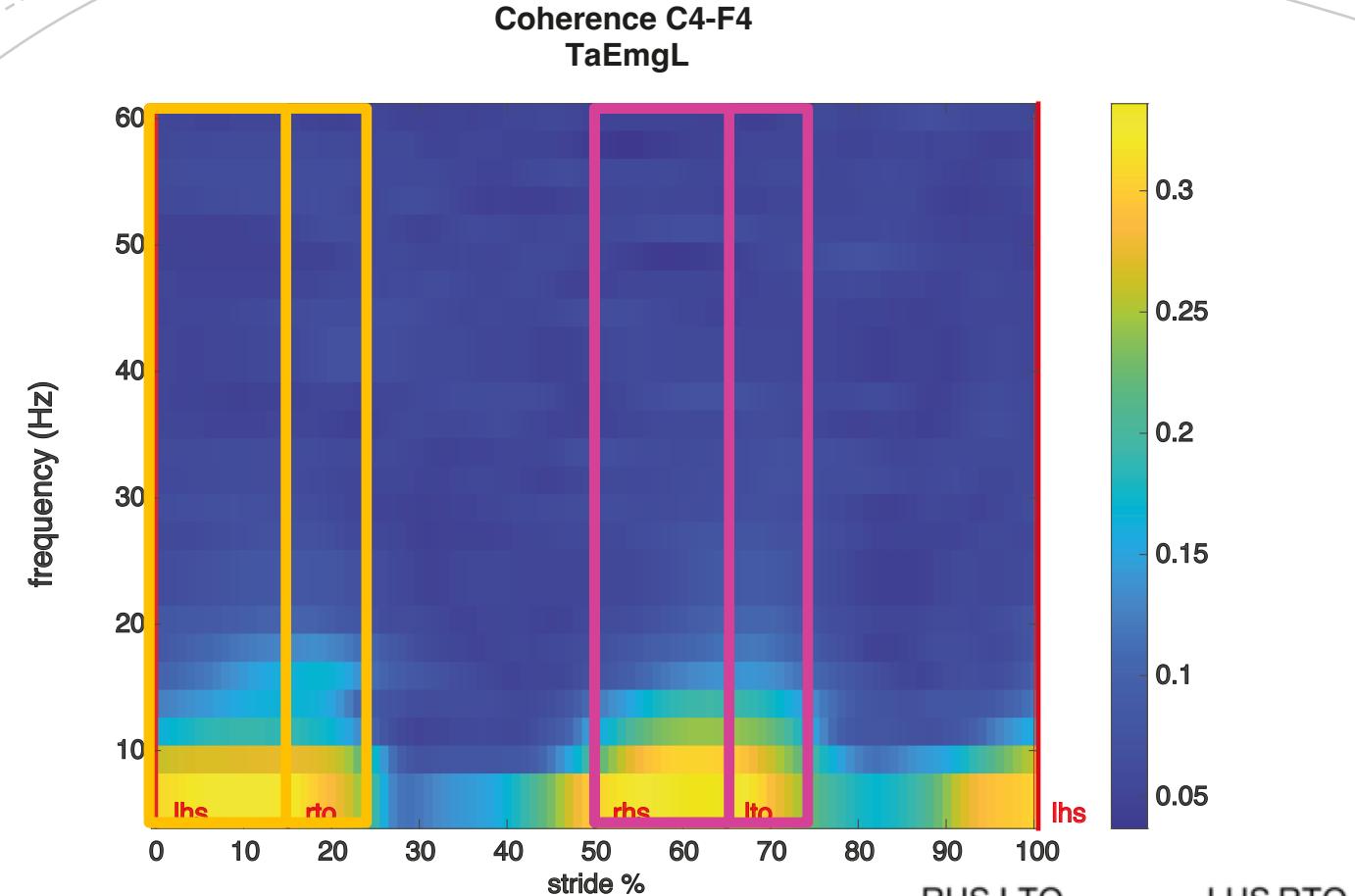
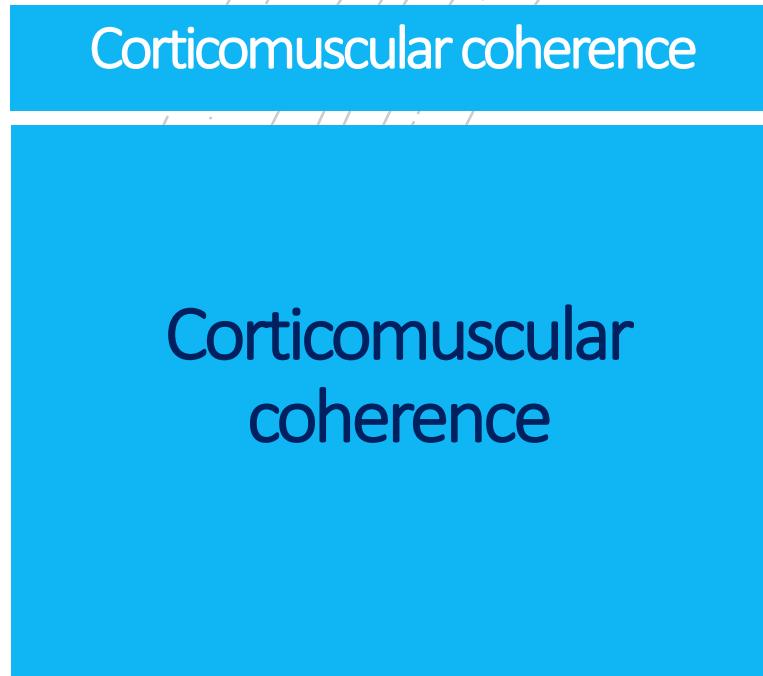


Tibialis Anterior C4- F4



Boonstra et al. 2020

ft_connectivityanalysis



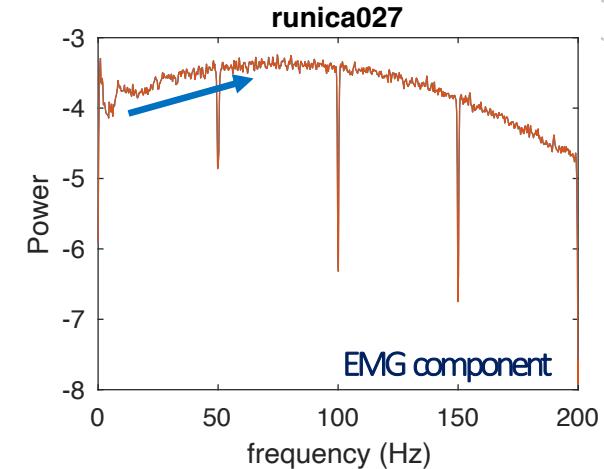
Corticomuscular coherence

Quick check:
EEG

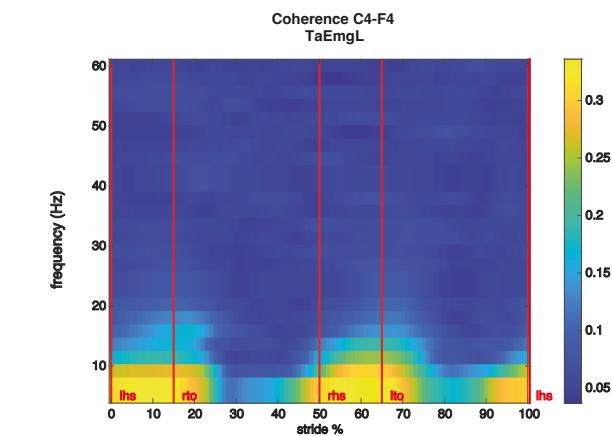
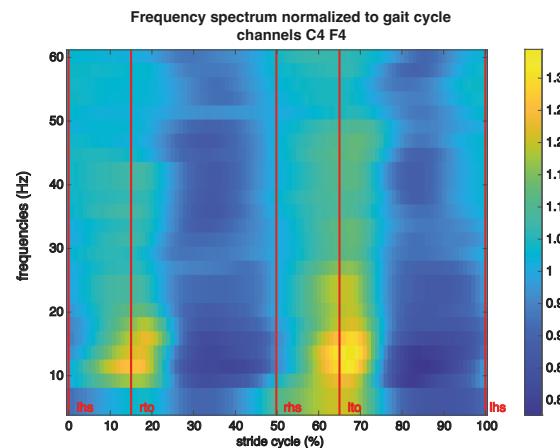
Long Quick check: EEG

Corticomuscular coherence

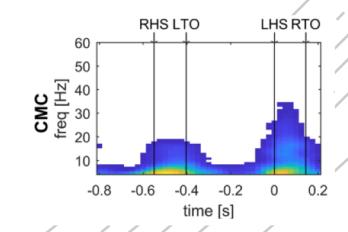
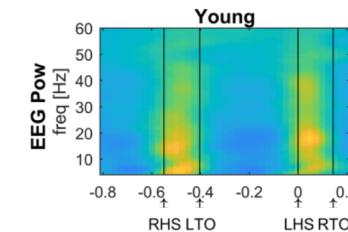
- Preprocessing
- ICA artefact removal



- Computing frequency power spectrum
- Computing coherence spectrum



- Comparing to the literature



Corticomuscular coherence

Next steps

- Improving data cleaning and preprocessing
- Computing results for muscles of interest
- Source localization
- Statistics

Corticomuscular coherence

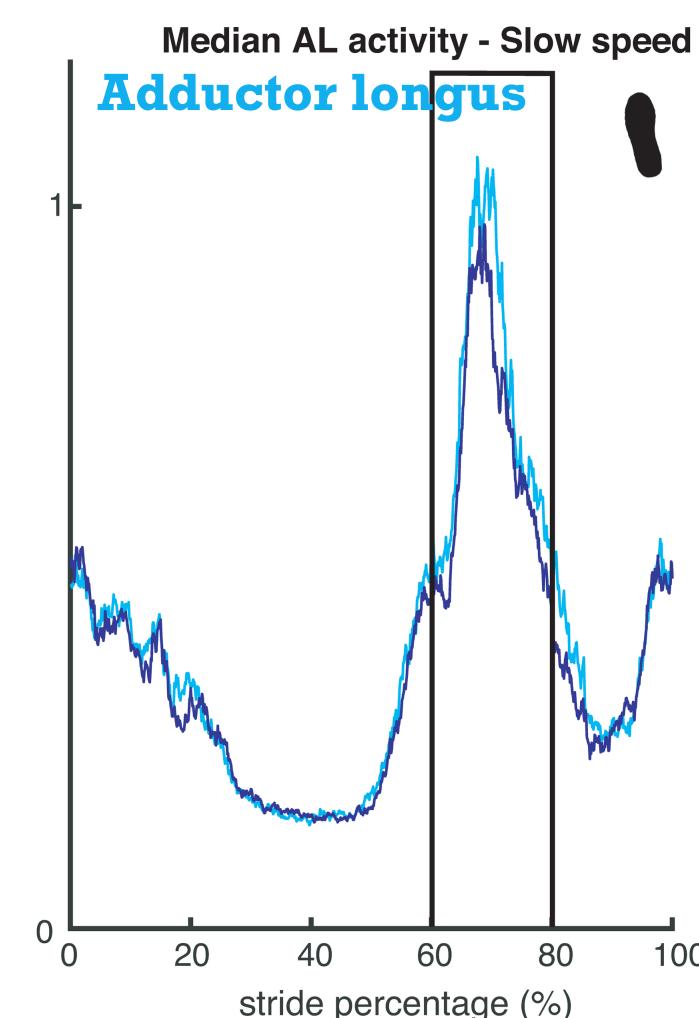
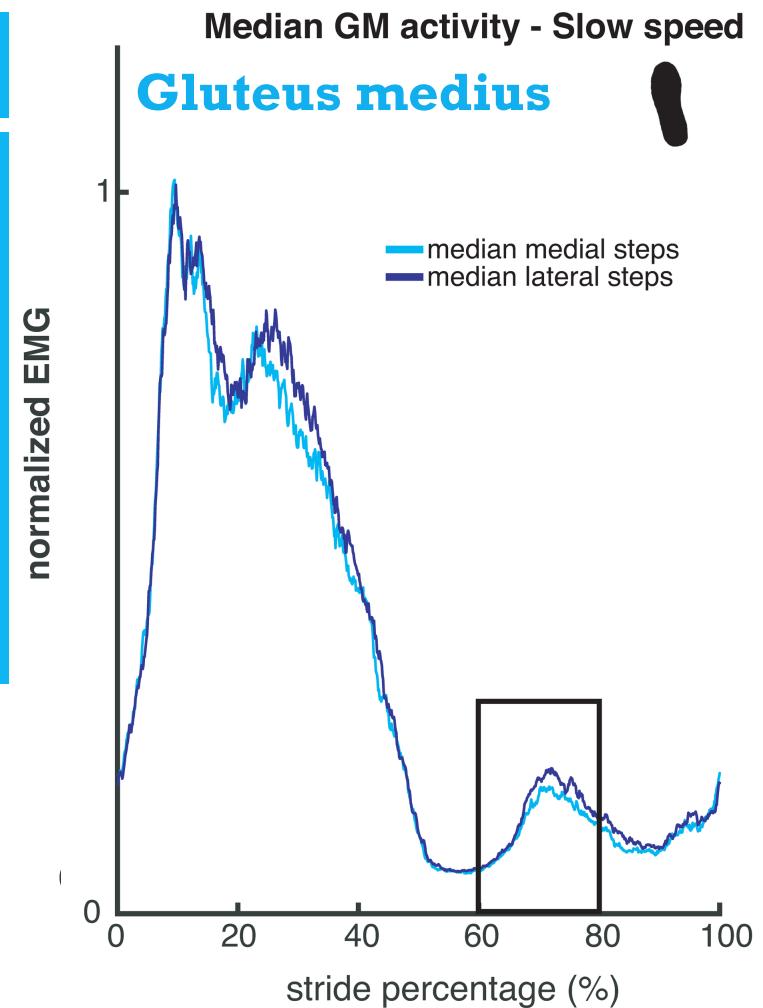
Next steps

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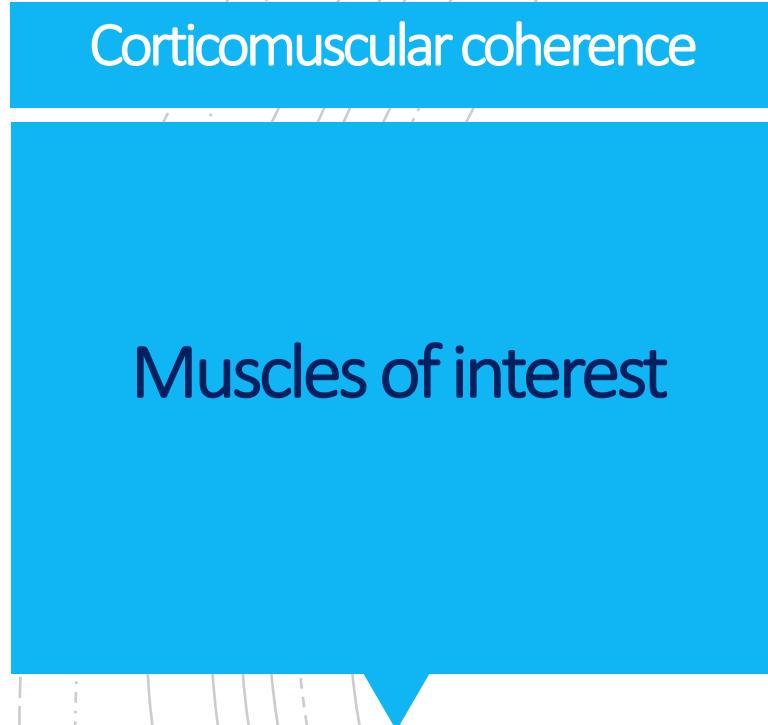
ML



$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

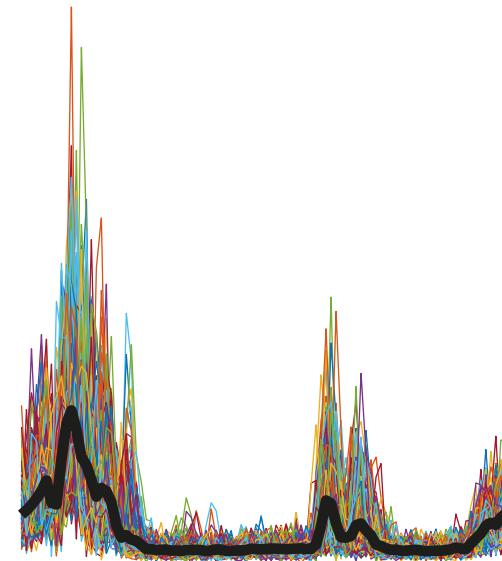


AP

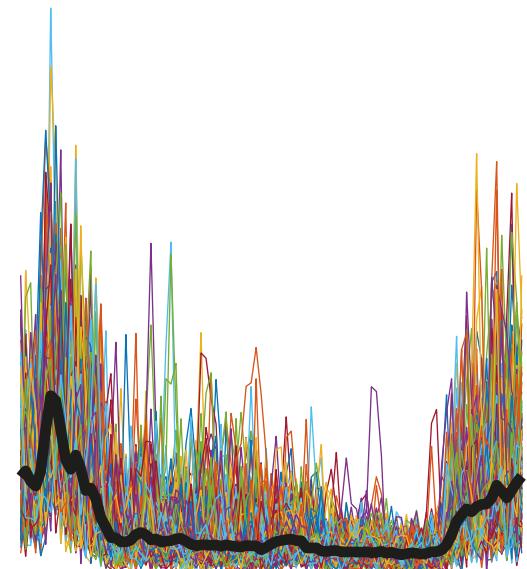


$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

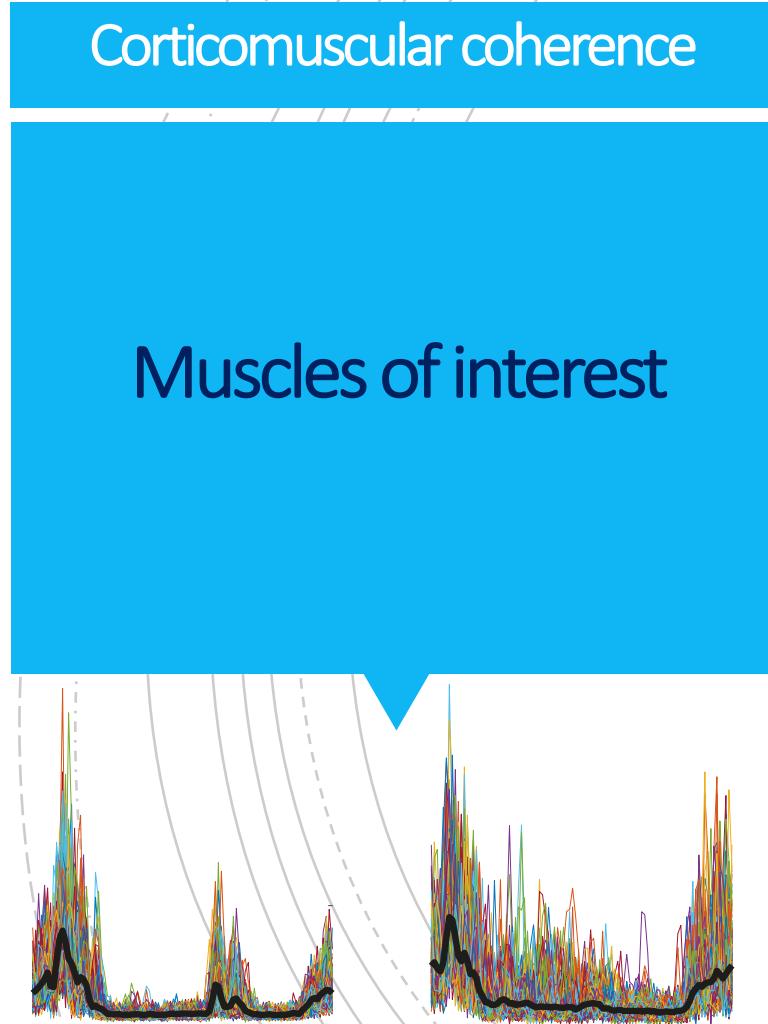
Rectus femoris



Biceps femoris



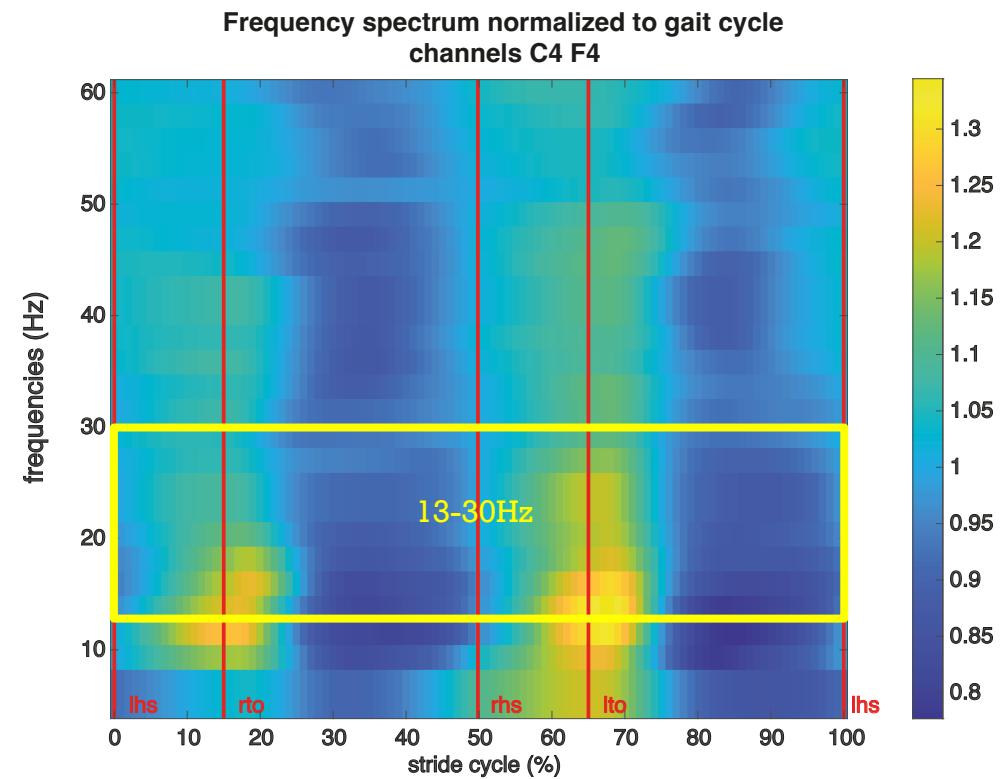
AP



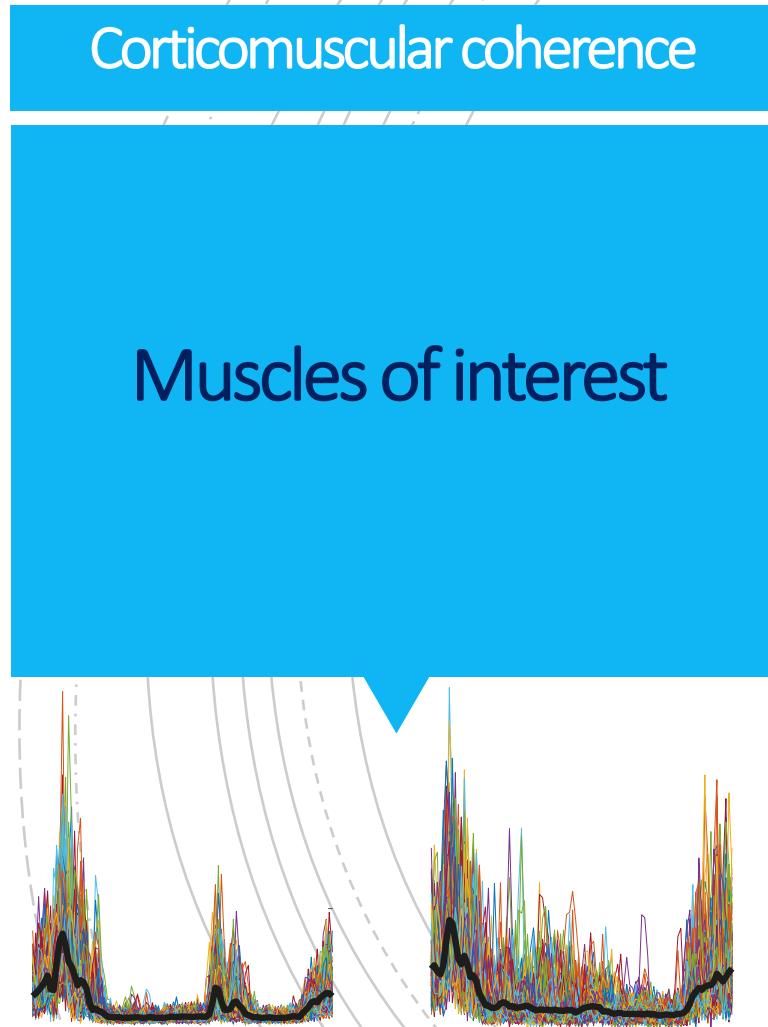
$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

Rectus femoris

Biceps femoris



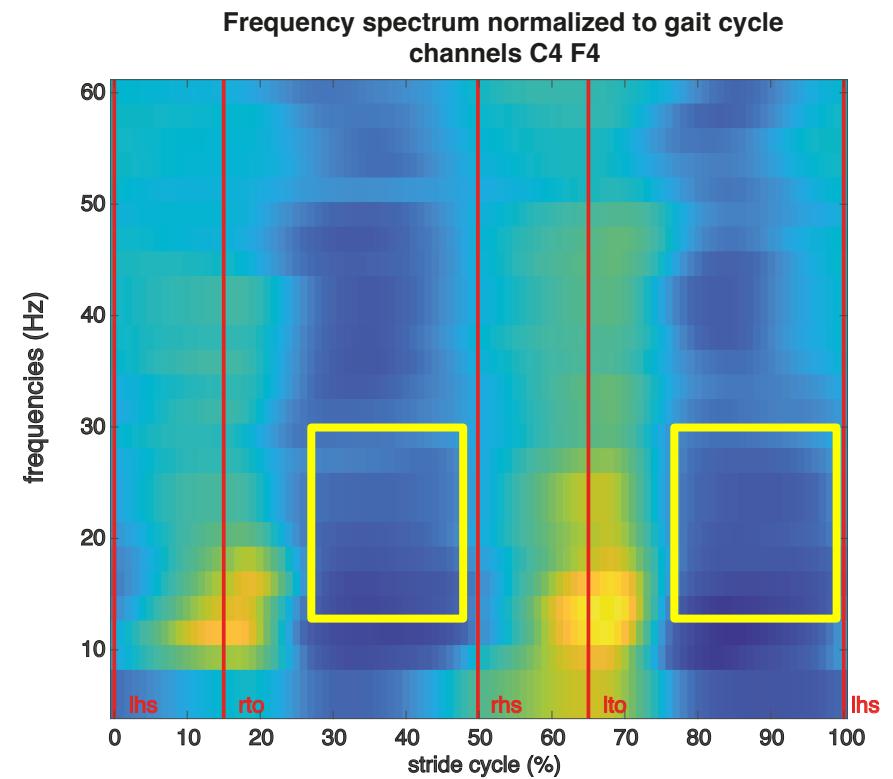
AP



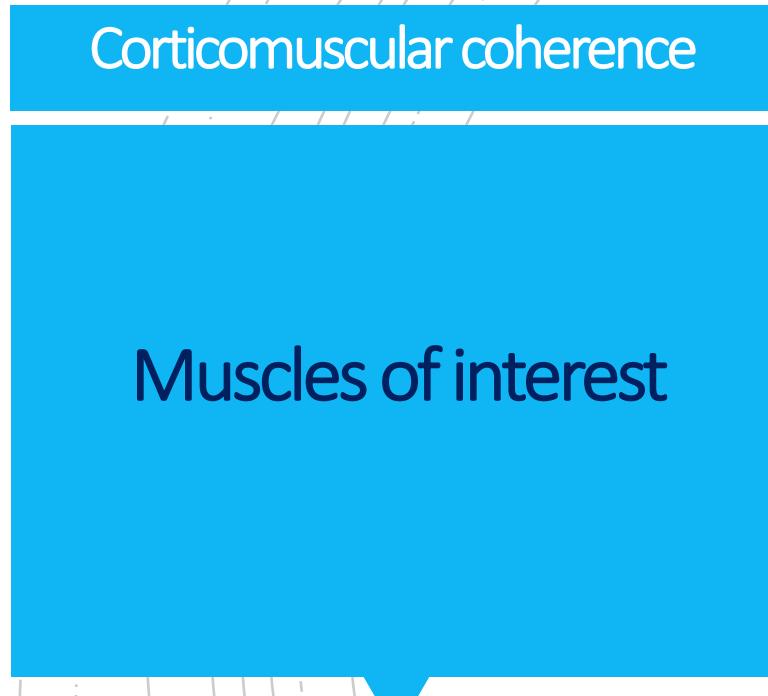
$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$

Rectus femoris

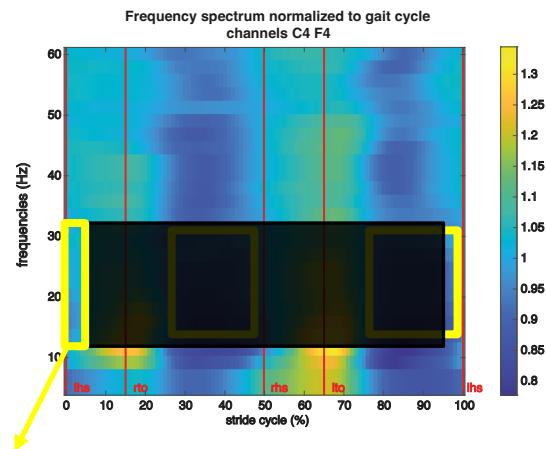
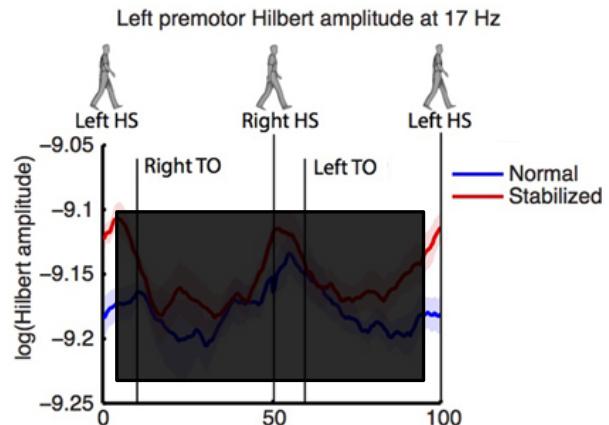
Biceps femoris



AP

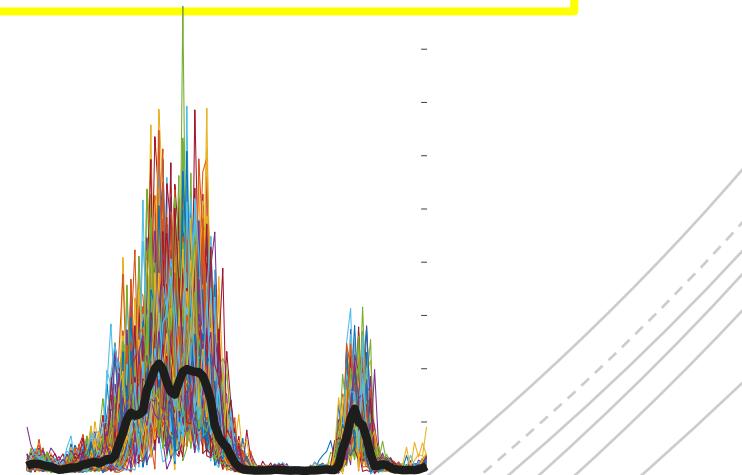


$$FP = \beta_{CoM_pos}(i) * CoM_{pos}(i) + \beta_{CoM_vel}(i) * CoM_{vel}(i) + \varepsilon$$



Double stance phase → preparing pushoff

Gastrocnemius



Corticomuscular coherence

Source localization

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Localizing oscillatory sources using beamformer technique

Introduction

In this tutorial we will show how to localise oscillatory sources using beamformer techniques. Below we will repeat the steps from the preprocessing tutorials (trigger-based segmentation and beamformer analysis).



ed in the preprocessing tutorials.
e data as described in the earlier

ends on:
n model

Quality sources

- Accuracy
- Electrod

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Localizing oscillatory sources using a 3D-scanner

Introduction

This tutorial shows how to create an electrode model based on a single subject's 3D scan. This electrode model can be used in combination with a **BEM** or **FEM** volume conduction model for source reconstruction.

Amsterdam team: prof. dr. Jaap van Dieën, dr. Sjoerd Bruijn, dr. Bernadette van Wijk
dr. Moira van Leeuwen

Foot Placement Control & Corticomuscular Coherence

Concepts &
Analysis Pipeline

Amsterdam
Movement
Sciences

VU VRIJE
UNIVERSITEIT
AMSTERDAM

 Amsterdam UMC
University Medical Centers