

MATLAB codes

Some of the codes used to process and to analyse the signals acquired during the measurements on the treadmill, are reported. In particular:

- The codes used for the IMUs calibration
- The codes for the cycle events identification:
 - Gold standard identification (based on the poles force signals)
 - Acceleration norm-based method
 - Angular velocity norm-based method
 - Fusion method

Main calibration

```
% main_calibration legge i dati dagli IMUs, sincronizza i sensori,  
% effettua la calibrazione secondo Ferraris 1995 -->  
% - 6 misure statiche in cui ogni asse del sensore è messo una volta  
% concorde ed una discorda a g. le 6 misure statiche devono essere  
% effettuate prima e dopo l'esperimento.  
% - ottengo i coefficienti correttivi che tengono conto dei vari errori  
% di misura (bias di acc e gyr, acc sensitivity del gyr, acc scale factor e  
% orientation matrix).  
  
% leggo i segnali dai sensori per ciascuna posizione statica sia per le misure  
% fatte prima che dopo l'esperimento.  
% per identificare la posizione statica -->  
% - la prima lettera mi dice quale asse è allineato con g  
% - se questa lettera è seguita da un underscore ( _ ) allora l'asse punta  
% verso il basso altrimenti punta verso l'alto  
% - la seconda lettera, che è seguita da 'eo' mi dice quale asse è  
% all'incirca in direzione est-ovest, info utile per la definizione  
% dell acc sensitivity del gyr.  
% - esempio1: ZYeo = asse Z verso l'alto e Y in direzione est-ovest  
% - esempio2: X_Yeo = asse X punta verso il basso, Y in direzione est-ovest  
  
% dopo aver effettuato le 6 misure prima e dopo, organizzo le prime 6 in una  
% cartella CalibBefore ed una cartella CalibAfter. Ognuna delle 6 misure  
% deve poi contenere nel suo nome l'identificativo della posizione assunta dal  
% sensore come specificato prima, questo è necessario per il riconoscimento  
% automatico della cartella esatta per ogni posizione da parte del codice  
% seguente. Le sei posizioni e quindi i 6 nomi identificativi sono ZYeo,  
% XYeo, Z_Yeo, X_Yeo, YXeo, Y_Zeo  
  
function [IMUsSignals_calib,calibParameters]=main_calibration(IMUsSignals)  
  
% valore di g. Inserire il valore esatto di g del posto in cui ci si trova.  
% si puo usare local gravity calculator online inserendo latitudine e altezza.  
g = 9.82172;  
  
%% data Reading  
  
elaborazione_Path = cd('..'); % Identify the script path  
cd .. % It moves one level up  
  
disp( 'select folder mem-IMUsCalib');
```

```

subjects_Path = uigetdir('C:\Users\alexa\Documents'); % mi apre un box dove ho la
cartella dei dati salvati
cd(subjects_Path) % mi muovo in quella cartella

subjectsLIST=dir('*subject*'); % leggo i folders dei
diversi subjects in una unica cartella
for h=1:length(subjectsLIST)

    subjectname=string(subjectsLIST(h).name);
    cd(subjectname)
    before_after_Path = cd('.');

    % salvo i nomi delle misure Before e After
    list=dir('*Before*');
    before_after(1)=string(list.name);
    list=dir('*After*');
    before_after(2)=string(list.name);

    for j=1:length(before_after)
        data_Path= fullfile(before_after_Path,char(before_after(j)));
        cd (data_Path)

        % salvo i nomi di ciascuna cartella corrispondente a ciascuna delle 6
        % misure statiche
        list=dir('*ZYeo*');
        PositionsFolders(1)=string(list.name); % nome Folder ZYeo
        list=dir('*XYeo*');
        PositionsFolders(2)=string(list.name); % nome Folder XYeo
        list=dir('*Z_Yeo*');
        PositionsFolders(3)=string(list.name); % nome Folder Z_Yeo
        list=dir('*X_Yeo*');
        PositionsFolders(4)=string(list.name); % nome Folder X_Yeo
        list=dir('*YXeo*');
        PositionsFolders(5)=string(list.name); % nome Folder YXeo
        list=dir('*Y_Zeo*');
        PositionsFolders(6)=string(list.name); % nome Folder Y_Zeo

        %ciclo for che itera sulle 6 misure statiche
        for k=1:6

            FolderPath= fullfile(data_Path,char(PositionsFolders(k)));
            cd (FolderPath)

            % faccio una lista, assegno un nome a ogni file
            LIST=dir('*.*csv*'); % leggo i file csv in
            un'unica cartella
            for i=1:length(LIST)

                sensorname(i)=string(LIST(i).name); % assegno nome a ogni
                sensore

            end
            cd(elaborazione_Path)

            %leggo i dati dai sensori con la function readsensor, per ogni
            sensore

```

```

        %ottengo una matrice con [time acc_x acc_y acc_z omega_x omega_y
omega_z]
        % notare che l'ordine dei sensori in sensorsraw corrisponde a quello
in
        % sensorname --> controllare in sensorname per sapere esattamente
quale
        % sensore si sta guardando in sensorsraw.
        for i=1:length(LIST)

[sensorsraw(i).sensor]=readSensor(g,fullfile(FolderPath,char(sensorname(i))));

        end
        % sincronizzazione dei sensori
        [calib(j).measurement(k).position]= synchronizeSensors(sensorsraw);

    end

        % vettore di due elementi contenete l'ora della prima e della seconda
        % misura di calibrazione
        t(j) = sensorsraw(1).sensor(1,1);
end

% time passed between two measurements in ms
tpassed = (t(2)-t(1));

% riorganizzo i dati in una struttura più comprensibile
for j=1:2
    calibPositions(j).ZYeo=calib(j).measurement(1).position;
    calibPositions(j).XYeo=calib(j).measurement(2).position;
    calibPositions(j).Z_Yeo=calib(j).measurement(3).position;
    calibPositions(j).X_Yeo=calib(j).measurement(4).position;
    calibPositions(j).YXeo=calib(j).measurement(5).position;
    calibPositions(j).Y_Zeo=calib(j).measurement(6).position;
end

%% controllo visivo segnali

% plot accelerazioni e vel ang di un solo sensore per la prima e la seconda
misura
% per essere sicuri di aver eseguito le misure statiche correttamente.

% for i=1:2
%     figure();
%     subplot(3,2,1)
%     plot(calibPositions(i).ZYeo(1).sensor.acc); ylabel({'acc (g)'}); ylim([-1.5 1.5]); legend('x','y','z'); title (['ZYeo misura',num2str(i)])
%     subplot(3,2,2)
%     plot(calibPositions(i).XYeo(1).sensor.acc); ylabel({'acc (g)'}); ylim([-1.5 1.5]); legend('x','y','z'); title (['XYeo misura',num2str(i)])
%     subplot(3,2,3)
%     plot(calibPositions(i).Z_Yeo(1).sensor.acc); ylabel({'acc (g)'}); ylim([-1.5 1.5]); legend('x','y','z'); title ( ['-ZYeo misura',num2str(i)])
%     subplot(3,2,4)
%     plot(calibPositions(i).X_Yeo(1).sensor.acc); ylabel({'acc (g)'}); ylim([-1.5 1.5]); legend('x','y','z'); title ( ['-XYeo misura',num2str(i)])
%     subplot(3,2,5)

```

```

% plot(calibPositions(i).YXeo(1).sensor.acc); ylabel({'acc (g)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['YXeo misura',num2str(i)])
% subplot(3,2,6)
% plot(calibPositions(i).Y_Zeo(1).sensor.acc); ylabel({'acc (g)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['-YZeo misura',num2str(i)])
% end
%
% for i=1:2
% figure();
% subplot(3,2,1)
% plot(calibPositions(i).ZYeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['ZYeo misura',num2str(i)])
% subplot(3,2,2)
% plot(calibPositions(i).XYeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['XYeo misura',num2str(i)])
% subplot(3,2,3)
% plot(calibPositions(i).Z_Yeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['-ZYeo misura',num2str(i)])
% subplot(3,2,4)
% plot(calibPositions(i).X_Yeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['-XYeo misura',num2str(i)])
% subplot(3,2,5)
% plot(calibPositions(i).YXeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['YXeo misura',num2str(i)])
% subplot(3,2,6)
% plot(calibPositions(i).Y_Zeo(1).sensor.gyr); ylabel({'vel ang (dps)'}) ; ylim([-1.5 1.5]); legend('x','y','z'); title (['-YZeo misura',num2str(i)])
% end

%% calibration parameters
for i=1:length(LIST)

[calibParameters(h).subject(i).sensor]=calibration(g,tpassed,calibPositions(1).ZYeo(i).sensor,calibPositions(1).XYeo(i).sensor,calibPositions(1).Z_Yeo(i).sensor,calibPositions(1).X_Yeo(i).sensor,calibPositions(1).YXeo(i).sensor,calibPositions(1).Y_Zeo(i).sensor,calibPositions(2).ZYeo(i).sensor,calibPositions(2).XYeo(i).sensor,calibPositions(2).Z_Yeo(i).sensor,calibPositions(2).X_Yeo(i).sensor,calibPositions(2).YXeo(i).sensor,calibPositions(2).Y_Zeo(i).sensor);
end
IMUsSignals_calib_tmp=[];
for i=1:length(IMUsSignals(h).subject)
    for j=1:length(IMUsSignals(h).subject(i).trial)
        % riscrivo il vettore dei tempi in relazione al tempo della prima calibrazione

IMUsSignals_calib_tmp(i).trial(j).sensor.time=IMUsSignals(h).subject(i).trial(j).sensor.time-t(1);
        for k=1:length(IMUsSignals(h).subject(i).trial(j).sensor.acc)

            IMUsSignals_calib_tmp(i).trial(j).sensor.acc(k,:)=
inv(calibParameters(h).subject(j).sensor.accOrient)*inv(calibParameters(h).subject(j).sensor.accScale)*(IMUsSignals(h).subject(i).trial(j).sensor.acc(k,:)')-
diag(calibParameters(h).subject(j).sensor.accBias));
            IMUsSignals_calib_tmp(i).trial(j).sensor.gyr(k,:)=
IMUsSignals(h).subject(i).trial(j).sensor.gyr(k,:)-
calibParameters(h).subject(j).sensor.gyroAccSens *
IMUsSignals(h).subject(i).trial(j).sensor.acc(k,:)'
-
```

```

calibParameters(h).subject(j).sensor.gyroBias0' +
calibParameters(h).subject(j).sensor.gyroBias1'
*IMUsSignals(h).subject(i).trial(j).sensor.time(k);

    %usare queste due righe sotto al posto di quelle sopra se non
    %si vuole correggere acc misalligment e gyro acc sensitivity
    %
    IMUsSignals_calib_tmp(i).trial(j).sensor.acc(k,:) =
inv(calibParameters(h).subject(j).sensor.accScale)*(IMUsSignals(h).subject(i).tri-
al(j).sensor.acc(k,:)' - diag(calibParameters(h).subject(j).sensor.accBias));
    %
    IMUsSignals_calib_tmp(i).trial(j).sensor.gyr(k,:) =
IMUsSignals(h).subject(i).trial(j).sensor.gyr(k,:)' -
calibParameters(h).subject(j).sensor.gyroBias0' +
calibParameters(h).subject(j).sensor.gyroBias1'
*IMUsSignals(h).subject(i).trial(j).sensor.time(k);

    end
end
end

% calcolo norma dell'acc e della vel ang per ogni sensore

for j=1:length(IMUsSignals_calib_tmp)
    for k=1:length(IMUsSignals_calib_tmp(j).trial)

        for i=1:length(IMUsSignals_calib_tmp(j).trial(k).sensor.acc(:,1))

            IMUsSignals_calib_tmp(j).trial(k).sensor.accNorm(i,1) =
norm(IMUsSignals_calib_tmp(j).trial(k).sensor.acc(i,:));
            IMUsSignals_calib_tmp(j).trial(k).sensor.gyrNorm(i,1) =
norm(IMUsSignals_calib_tmp(j).trial(k).sensor.gyr(i,:));

        end
    end
end

end
IMUsSignals_calib(h).subject=IMUsSignals_calib_tmp;
cd(subjects_Path)
end
cd(elaborazione_Path)
end

```

Function called in main calibration

```

% calibration---> calculation of the parameters for the correction of the
%                   accelerometer offset, sensitivity and misallignment
%                   and of the gyro offset and acceleration sensitivity
%
% input--> - g = valore locale dell'acc gravitazionale
%           - dt = tempo trascorso tra la prima e la seconda misura di
%                 calibrazione ( quella fatta prima e quella dopo l'esperimento)
%           - le 6 misure statiche fatte prima dell'esperimento
%           - le 6 misure statiche fatte dopo l'esperimento
%
% output--> - calibParameters = struttura contenente tutti i parametri utili
%               alla correzione degli errori

```

```

function [calibParameters] =
calibration(g,dt,ZYeo,XYeo,Z_Yeo,X_Yeo,YXeo,Y_Zeo,ZYeoAfter,XYeoAfter,Z_YeoAfter,
X_YeoAfter,YXeoAfter,Y_ZeoAfter)

g=1; %perchè sto passando già i segnali espressi in g!! altrimenti mi viene
tutto moltiplicato di un fattore g
    % nel caso si passino i segnali non espressi in g allora togliere questa
riga

% calcolo la media nel tempo per ogni posizione e per ogni canale di acc e gyr
% ottengo dei vettori riga, ogni colonna rapp. output x,y,z.
ZYeo.acc = mean(ZYeo.acc);
XYeo.acc = mean(XYeo.acc);
Z_Yeo.acc = mean(Z_Yeo.acc);
X_Yeo.acc = mean(X_Yeo.acc);
YXeo.acc = mean(YXeo.acc);
Y_Zeo.acc = mean(Y_Zeo.acc);
ZYeo.gyr = mean(ZYeo.gyr);
XYeo.gyr = mean(XYeo.gyr);
Z_Yeo.gyr = mean(Z_Yeo.gyr);
X_Yeo.gyr = mean(X_Yeo.gyr);
YXeo.gyr = mean(YXeo.gyr);
Y_Zeo.gyr = mean(Y_Zeo.gyr);
ZYeoAfter.acc = mean(ZYeoAfter.acc);
XYeoAfter.acc = mean(XYeoAfter.acc);
Z_YeoAfter.acc = mean(Z_YeoAfter.acc);
X_YeoAfter.acc = mean(X_YeoAfter.acc);
YXeoAfter.acc = mean(YXeoAfter.acc);
Y_ZeoAfter.acc = mean(Y_ZeoAfter.acc);
ZYeoAfter.gyr = mean(ZYeoAfter.gyr);
XYeoAfter.gyr = mean(XYeoAfter.gyr);
Z_YeoAfter.gyr = mean(Z_YeoAfter.gyr);
X_YeoAfter.gyr = mean(X_YeoAfter.gyr);
YXeoAfter.gyr = mean(YXeoAfter.gyr);
Y_ZeoAfter.gyr = mean(Y_ZeoAfter.gyr);

%% acc BIAS

Ua = cat(2, XYeo.acc', YXeo.acc', ZYeo.acc');
Ua_ = cat(2, X_Yeo.acc', Y_Zeo.acc', Z_Yeo.acc');
Uas = (Ua + Ua_);
Ba = Uas./2;

calibParameters.accBias = Ba;
%% acc SCALE FACTORS and ORIENTATION MATRIX

Uad = (Ua-Ua_);
Ka = diag(sqrt(diag(Uad*Uad')) ./ (4*(g)^2)));
Ra = (inv(Ka).* (1/(2*g))*Uad);

calibParameters.accScale = Ka;
calibParameters.accOrient = Ra;
%% gyro BIAS

Bg0 = [YXeo.gyr(1) XYeo.gyr(2) Y_Zeo.gyr(3)];
BgAfter = [YXeoAfter.gyr(1) XYeoAfter.gyr(2) Y_ZeoAfter.gyr(3)];
Bg1 = (BgAfter-Bg0)/dt;

```

```

calibParameters.gyroBias0 = Bg0;
calibParameters.gyroBias1 = Bg1;

%% gyro ACCELERATION SENSITIVITY

Ugp = [XYeo.gyr(1) YXeo.gyr(1) ZYeo.gyr(1)];
Uga = [X_Yeo.gyr(1) Y_Zeo.gyr(1) Z_Yeo.gyr(1)];
Kga = (Ugp-Uga)/(2*g);
Kga = diag(Kga);
calibParameters.gyroAccSens = Kga;
end

```

Cycle events identification from the force signals (gold standard)

```

% identification Hits and Leaves from Coachtech

% identifica gli istanti nel segnali di forza da coachtech in cui i
% bastoncini colpiscono e lasciano il terreno all'interno di una finestra
% definita.

% input: -CoachtechSignals_sinc --> segnali da coachtech, sincronizzati
%        -startSample --> campione che definisce l'inizio dell'intervallo di
%                           osservazione

% output: -Coach_pole_events --> struttura che contiene gli istanti di
%           contatto e distacco tra bastoncini e terreno per tutti i trial
%           considerati, calcolati a partire dal segnale di F di coachtech

function [Coach_pole_events] =
Hits_Leaves_Coachtech(CoachtechSignals_sinc,startSample)

    % identificazione poleLeaves
    for i=1:length(CoachtechSignals_sinc)

        [pks,locs,w,p] =
findpeaks(ribalta(CoachtechSignals_sinc(i).forceR(startSample(i):end)), 'MinPeakHeight', 0.90*max(ribalta(CoachtechSignals_sinc(i).forceR(startSample(i):end))), 'MinPeakDistance', 340, 'MinPeakProminence', 4);
        figure()
        %
        findpeaks(ribalta(CoachtechSignals_sinc(i).forceR(startSample(i):end)), 'MinPeakHeight', 0.90*max(ribalta(CoachtechSignals_sinc(i).forceR(startSample(i):end))), 'MinPeakDistance', 340, 'MinPeakProminence', 4);
        Coach_pole_events(i).Leaves.sample = locs + startSample(i)-1;
        Coach_pole_events(i).Leaves.peak=
CoachtechSignals_sinc(i).forceR(Coach_pole_events(i).Leaves.sample);

    end

    % identificazione poleHits
    for i=1:length(CoachtechSignals_sinc)

        for j=1:length(Coach_pole_events(i).Leaves.sample)-1

            [pks,locs,w,p] =
findpeaks(CoachtechSignals_sinc(i).forceR(Coach_pole_events(i).Leaves.sample(j,1))

```

```

:Coach_pole_events(i).Leaves.sample(j+1,1)), 'MinPeakHeight', 0.2*max(CoachtechSign
als_sinc(i).forceR(Coach_pole_events(i).Leaves.sample(j,1):Coach_pole_events(i).L
eaves.sample(j+1,1)));
%
figure()
%
findpeaks(CoachtechSignals_sinc(i).forceR(Coach_pole_events(i).Leaves.sample(j,1)
:Coach_pole_events(i).Leaves.sample(j+1,1)), 'MinPeakHeight', 0.2*max(CoachtechSign
als_sinc(i).forceR(Coach_pole_events(i).Leaves.sample(j,1):Coach_pole_events(i).L
eaves.sample(j+1,1)));
    Coach_pole_events(i).Hits.sample(j,1)=
locs(1)+Coach_pole_events(i).Leaves.sample(j,1)-1;

end

Coach_pole_events(i).Hits.peak=
CoachtechSignals_sinc(i).forceR(Coach_pole_events(i).Hits.sample);
%
figure()
%
ax1=subplot(3,1,1);
plot(ax1,CcoachtechSignals_sinc(i).forceR)
hold on
%
plot(ax1,Ccoach_pole_events(i).Hits.sample,Ccoach_pole_events(i).Hits.peak,'o',Coac
h_pole_events(i).Leaves.sample,Ccoach_pole_events(i).Leaves.peak,'square' )
%
grid on
%
ax2=subplot(3,1,2);
plot(ax2,pulse_sensor_sinc(i).trial.accNorm)
hold on
%
plot(ax2,poleHits(i).trial,peakHits(i).trial,'o',poleLeaves(i).trial,peakLeaves
(i).trial,'square' )
%
grid on
%
ax3=subplot(3,1,3);
plot(ax3,CcoachtechSignals_sinc(i).speed); ylabel({'Speed (km/h)'});
%
hold on
%
plot(ax3,smoothdata(CcoachtechSignals_sinc(i).speed,'movmean',30));
%
grid on
linkaxes([ax1 ax2 ax3],'x')
end
end

```

Function to refine the cycle events identification from the force signals

```

% identification Hits and Leaves from Coachtech

% identifica gli istanti nel segnali di forza da coachtech in cui i
% bastoncini colpiscono e lasciano il terreno all'interno di una finestra
% definita.

% input: -CoachtechSignals_sinc --> segnali da coachtech, sincronizzati
%        -startSample --> campione che definisce l'inizio dell'intervallo di
%                          osservazione

% output: -Coach_pole_events --> struttura che contiene gli istanti di
%           contatto e distacco tra bastoncini e terreno per tutti i trial
%           considerati, calcolati a partire dal segnale di F di coachtech

```

```

function [Coach_pole_events_refined, CoachtechSignals_sinc_unbiased] =
Hits_Leaves_Coachtech_refine(CoachtechSignals_sinc, Coach_pole_events)

    CoachtechSignals_sinc_unbiased=CoachtechSignals_sinc;

    window1=85; % campioni di cui mi sposto da hits vecchi
    window2=10;% campioni di cui mi sposto da hits nuovi e vecchi

    % smooth di force signal cosi le parti in cui i bastoncini non sono a
    contatto e
    % dovrebbero perciò essere zero rimangono più costanti

    for i=1:length(CoachtechSignals_sinc)

        CoachtechSignals_smoothed(i).forceR=smooth(CoachtechSignals_sinc(i).forceR,12);
        %
        figure()
        %
        plot(CoachtechSignals_sinc(i).forceR)
        %
        hold on
        %
        plot(CoachtechSignals_smoothed(i).forceR)
    end

    for i=1:length(CoachtechSignals_smoothed)
        asintoto=[];
        for j=1:length(Coach_pole_events(i).Hits.sample)

            dev(i).samples(j).cycle=Coach_pole_events(i).Hits.sample(j,1)-
            window1:Coach_pole_events(i).Hits.sample(j,1)-1;

            dev(i).values(j).cycle=diff(CoachtechSignals_smoothed(i).forceR(Coach_pole_events
            (i).Hits.sample(j,1)-window1:Coach_pole_events(i).Hits.sample(j,1)));

            outliers=dev(i).samples(j).cycle(isoutlier(dev(i).values(j).cycle,'percentiles',[0,99]));
            Coach_pole_events_refined(i).Hits.sample(j,1)=outliers(1);

            asintoto= [asintoto;
            CoachtechSignals_smoothed(i).forceR(dev(i).samples(j).cycle(1):dev(i).samples(j).
            cycle(end-window2))];
        end

        bias(i)=mean(asintoto);

        CoachtechSignals_sinc_unbiased(i).forceR=CoachtechSignals_sinc(i).forceR-
        bias(i);

        for j=1:length(Coach_pole_events_refined(i).Hits.sample)
            tmp=
            find(CoachtechSignals_sinc_unbiased(i).forceR(Coach_pole_events_refined(i).Hits.s
            ample(j):Coach_pole_events(i).Leaves.sample(j)) < 0.6) ;

            Coach_pole_events_refined(i).Leaves.sample(j,1)=Coach_pole_events_refined(i).Hits.
            sample(j,1)+tmp(1)-1;
        end
    end
end

```

```

Coach_pole_events_refined(i).Leaves.peak=CoachtechSignals_sinc_unbiased(i).forceR
(Coach_pole_events_refined(i).Leaves.sample);

Coach_pole_events_refined(i).Hits.peak=CoachtechSignals_sinc_unbiased(i).forceR(C
oach_pole_events_refined(i).Hits.sample);

end

% for i=1:length(CoachtechSignals_sinc_unbiased)
%
% figure()
% ax1=subplot(3,1,1);
% plot(ax1,CoachtechSignals_sinc_unbiased(i).forceR)
% hold on
%
plot(ax1,Coach_pole_events(i).Hits.sample,Coach_pole_events(i).Hits.peak,'o',Coac
h_pole_events(i).Leaves.sample,Coach_pole_events(i).Leaves.peak,'square' )
%
for j=1:length(Coach_pole_events(i).Hits.sample)-1
%
hold on
%
plot(ax1,dev(i).samples(j).cycle,dev(i).values(j).cycle,'r')
%
end
%
hold on
%
plot(ax1,Coach_pole_events_refined(i).Hits.sample,Coach_pole_events_refined(i).Hi
ts.peak,'*')
%
hold on
%
plot(ax1,Coach_pole_events_refined(i).Leaves.sample,Coach_pole_events_refined(i).L
eaves.peak,'X')
%
grid on
ax2=subplot(3,1,2);
plot(ax2,pulse_sensor_sinc(i).trial.accNorm)
%
hold on
%
plot(ax2,poleHits(i).trial,peaksHits(i).trial,'o',poleLeaves(i).trial,peaksLeaves
(i).trial,'square' )
%
grid on
%
ax3=subplot(3,1,3);
plot(ax3,CoachtechSignals_sinc_unbiased(i).speed); ylabel({'Speed
(km/h)'});
%
hold on
%
plot(ax3,smoothdata(CoachtechSignals_sinc_unbiased(i).speed,'movmean',30));
%
grid on
%
linkaxes([ax1 ax2 ax3],'x')
%
end

end

```

Cycle events identification: acceleration norm-based method

```
% identification Hits and Leaves from linear acceleration norm signal

function [pole_events] =
Hits_Leaves_IMUs_improved_new(wrist_sensor_sinc,wrist_sensor_filt, ...
                                wrist_sensor_filt_high,startSample,Fs)

%% questa parte commentata usarla se si passa alla funzione la norma non filtrata
% ma no se si passa la norma high e low
%     %filtraggio high
%     fcH = 40; %cutting freq
%
%     [bH,aH] = butter(2,fcH/(Fs/2),'high');
%     % freqz(bH,aH,Fs,Fs)
%
%     fc = 4; %cutting freq
%
%     [b,a] = butter(2,fc/(Fs/2));
%     % freqz(b,a,Fs,Fs)
%
%     for i=1:length(wrist_sensor_sinc)
%
%         wrist_sensor_filt(i).trial=filtfilt(b,a,wrist_sensor_sinc(i).trial);
%
wrist_sensor_filt_high(i).trial=filtfilt(bH,aH,wrist_sensor_sinc(i).trial);
%
figure()
time=(0:length(wrist_sensor_sinc(i).trial)-1).*1/Fs;
plot(time,wrist_sensor_sinc(i).trial,'LineWidth',1.5,'Color','k')
hold on
plot(time,wrist_sensor_filt(i).trial,'LineWidth',1.5,'Color','r')
title ('accNorm raw signal VS filtered signal')
xlabel('time (s)')
ylabel('acceleration norm (g)')

figure()
plot(time,wrist_sensor_sinc(i).trial,'LineWidth',1.5,'Color','k')
hold on
plot(time,wrist_sensor_filt_high(i).trial,'LineWidth',1.5,'Color','r')
title ('accNorm raw signal VS filtered signal')
xlabel('time (s)')
ylabel('acceleration norm (g)')

end

%% voglio usare il segnale raw per identificazione lifts -- > migliori
% prestazioni
% se si volesse provare a farla con il segnale filtrato basta commentare la
% riga qui sotto e andare a fare un filtraggio appropriato nella funzione
% spectral analysis
wrist_sensor_filt=wrist_sensor_sinc;
%%

% identificazione polehits
for i=1:length(wrist_sensor_sinc)
```

```

[pks,locs,w,p] =
findpeaks(wrist_sensor_filt_high(i).trial(startSample(i):end), 'MinPeakHeight', 0.1
0*max(wrist_sensor_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance', 340)
;
%
figure()
t=(0:length(wrist_sensor_filt_high(i).trial(startSample(i):end))-1).*1/Fs;
%
findpeaks(wrist_sensor_filt_high(i).trial(startSample(i):end),t, 'MinPeakHeight', 0
.10*max(wrist_sensor_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance', 34
0*1/Fs);
%
title ('find peaks in accNorm filtered signal')
xlabel('time (s)')
ylabel('acceleration norm (g)')
pole_events(i).Hits.sample = locs + startSample(i)-1;
pole_events(i).Hits.peak=
wrist_sensor_sinc(i).trial(pole_events(i).Hits.sample);
end

% identificazione poleleaves
for i=1:length(wrist_sensor_filt)
    windows=round(diff(pole_events(i).Hits.sample)./10);

    for j=1:length(pole_events(i).Hits.sample)-1

        [pks,locs,w,p] =
findpeaks(wrist_sensor_filt(i).trial(pole_events(i).Hits.sample(j,1)+windows(j):p
ole_events(i).Hits.sample(j+1,1)-
windows(j)), 'MinPeakHeight', 0.90*max(wrist_sensor_filt(i).trial(pole_events(i).Hi
ts.sample(j,1)+windows(j):pole_events(i).Hits.sample(j+1,1)-
windows(j))), 'MinPeakDistance', 220);
%
        figure()

t=(0:length(wrist_sensor_filt(i).trial(pole_events(i).Hits.sample(j,1)+60:pole_eve
nts(i).Hits.sample(j+1,1)-60))-1).* (1/Fs).*1000;
%
findpeaks(wrist_sensor_filt(i).trial(pole_events(i).Hits.sample(j,1)+60:pole_eve
nts(i).Hits.sample(j+1,1)-
60),t, 'MinPeakHeight', 0.90*max(wrist_sensor_filt(i).trial(pole_events(i).Hits.sam
ple(j,1)+60:pole_events(i).Hits.sample(j+1,1)-
60)), 'MinPeakDistance', 220*(1/Fs)*1000);
%
        title ('window between two hits (filtered signal)')
xlabel('time (ms)')
ylabel('acceleration norm (g)')
if length(locs) > 1
    pole_events(i).Leaves.sample(j,1)= locs(end-
1)+pole_events(i).Hits.sample(j,1)+60-1;
elseif isempty(locs)==1
    pole_events(i).Leaves.sample(j,1)=
pole_events(i).Hits.sample(j)+round((pole_events(i).Hits.sample(j+1)-
pole_events(i).Hits.sample(j))./3);
else
    pole_events(i).Leaves.sample(j,1)=
locs(end)+pole_events(i).Hits.sample(j,1)+60-1;
end

if isempty(locs)==1

```

```

        pole_events(i).Leaves.sample(j,1)=
pole_events(i).Hits.sample(j)+round((pole_events(i).Hits.sample(j+1)-
pole_events(i).Hits.sample(j))./3);
    else
        pole_events(i).Leaves.sample(j,1)=
locs(1)+pole_events(i).Hits.sample(j,1)+windows(j)-1;
    end

end

pole_events(i).Leaves.peak=
wrist_sensor_sinc(i).trial(pole_events(i).Leaves.sample);

%
figure()
%
ax1=subplot(2,1,1);
%
plot(ax1,wrist_sensor_sinc(i).trial)
%
hold on
%
plot(ax1,pole_events(i).Hits.sample,pole_events(i).Hits.peak,'o',pole_events(i).L
eaves.sample,pole_events(i).Leaves.peak,'square' )
%
grid on
%
ax2=subplot(2,1,2);
%
plot(ax2,Coachtechwrist_sensor_sinc_sinc(i).speed); ylabel({'Speed
(km/h)'});
%
hold on
%
plot(ax2,smoothdata(Coachtechwrist_sensor_sinc_sinc(i).speed,'movmean',30));
%
grid on
%
linkaxes([ax1 ax2],'x')

end

%% pole hits refine
window1=5; % campioni di cui mi sposto dall'hit
window2=10; % finestra in campioni in cui identifico la variazione della
pendenza curva

% smooth di signal

for i=1:length(wrist_sensor_sinc)

    wrist_sensor_sinc_smoothed(i).trial=smooth(wrist_sensor_sinc(i).trial,5);
%
figure()
%
plot(wrist_sensor_sinc(i).trial)
%
hold on
%
plot(wrist_sensor_sinc_smoothed(i).trial)
end

for i=1:length(wrist_sensor_sinc_smoothed)

    for j=1:length(pole_events(i).Hits.sample)

        dev(i).samples(j).cycle=pole_events(i).Hits.sample(j,1)-
window2:pole_events(i).Hits.sample(j,1)-window1-1;
    end
end

```

```

dev(i).values(j).cycle=diff(wrist_sensor_sinc_smoothed(i).trial(pole_events(i).Hits.sample(j,1)-window2:pole_events(i).Hits.sample(j,1)-window1));

outliers=dev(i).samples(j).cycle(isoutlier(dev(i).values(j).cycle,'percentiles',[0,88]));
pole_events_refined(i).Hits.sample(j,1)=outliers(1);
end
pole_events_refined(i).Leaves.sample=pole_events(i).Leaves.sample;

pole_events_refined(i).Leaves.peak=wrist_sensor_sinc(i).trial(pole_events_refined(i).Leaves.sample);

pole_events_refined(i).Hits.peak=wrist_sensor_sinc(i).trial(pole_events_refined(i).Hits.sample);

end
pole_events= pole_events_refined;
% for i=1:length(wrist_sensor_sinc)
%
% figure()
% plot(wrist_sensor_sinc(i).trial)
% hold on
%
plot(pole_events(i).Hits.sample,pole_events(i).Hits.peak,'o',pole_events(i).Leaves.sample,pole_events(i).Leaves.peak,'square' )
% for j=1:length(pole_events(i).Hits.sample)-1
% hold on
% plot(dev(i).samples(j).cycle,dev(i).values(j).cycle,'r')
% end
% hold on
%
plot(pole_events_refined(i).Hits.sample,pole_events_refined(i).Hits.peak,'*')
% hold on
%
plot(pole_events_refined(i).Leaves.sample,pole_events_refined(i).Leaves.peak,'x')
%
end
end

```

Cycle events identification: angular velocity norm-based method

```

% identification Hits and Leaves from angular velocity norm signal

function
[pole_events_gyr]=Hits_Leaves_IMUs_gyr00(gyrNorm,gyrNorm_filt,gyrNorm_filt_high,startSample,Fs)
%% questa parte commentata usarla se si passa alla funzione la norma non filtrata
ma no se si passa la norma high e low
%filtraggio angular vel high
% fcH = 60; %cutting freq
%
% [bH,aH] = butter(2,fcH/(Fs/2),'high');
% % freqz(bH,aH,Fs,Fs)

```

```

%
% % filtraggio angular vel
%
% fc = 20; %cutting freq
%
% [b,a] = butter(2,fc/(Fs/2));
% % freqz(b,a,Fs,Fs)
%
% for i=1:length(gyrNorm)
%     gyrNorm_filt(i).trial=filtfilt(b,a,gyrNorm(i).trial);
%     gyrNorm_filt_high(i).trial=filtfilt(bH,aH,gyrNorm(i).trial);
%
%     figure()
%     time=(0:length(gyrNorm(i).trial)-1).* (1/Fs);
%     plot(time,gyrNorm(i).trial,'LineWidth',1.5,'Color','k')
%     hold on
%     plot(time,gyrNorm_filt(i).trial,'LineWidth',1.5,'Color','r')
%     title ('gyrNorm raw signal VS filtered signal')
%     xlabel('time (s)')
%     ylabel('angular velocity norm (dps)')
%
%     figure()
%     plot(time,gyrNorm(i).trial,'LineWidth',1.5,'Color','k')
%     hold on
%     plot(time,gyrNorm_filt_high(i).trial,'LineWidth',1.5,'Color','r')
%     title ('gyrNorm raw signal VS filtered signal')
%     xlabel('time (s)')
%     ylabel('angular velocity norm (dps)')
%
% end
%% voglio usare il segnale raw per identificazione lifts -- > migliori
prestazioni
% se si volesse provare a farla con il segnale filtrato basta commentare la
% riga qui sotto e andare a fare un filtraggio appropriato nella funzione
% spectral analysis
gyrNorm_filt=gyrNorm;
%%
% identificazione polehits
for i=1:length(gyrNorm)

    [pkss,locs,w,p] =
findpeaks(gyrNorm_filt_high(i).trial(startSample(i):end), 'MinPeakHeight',0.10*max
(gyrNorm_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance',350);
%
    figure()
%        t=(0:length(gyrNorm_filt_high(i).trial(startSample(i):end))-1).* (1/Fs);
%
findpeaks(gyrNorm_filt_high(i).trial(startSample(i):end),t, 'MinPeakHeight',0.10*m
ax(gyrNorm_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance',350*(1/Fs));
%
    title ('find peaks in gyrNorm filt signal')
    xlabel('time (s)')
    ylabel('angular vel (dps)')
    pole_events_gyr(i).Hits.sample = locs + startSample(i)-1;
    pole_events_gyr(i).Hits.peak=
gyrNorm(i).trial(pole_events_gyr(i).Hits.sample);
end

% identificazione poleLeaves

```

```

for i=1:length(gyrNorm)

    windows=round(diff(pole_events_gyr(i).Hits.sample)./10);

    for j=1:length(pole_events_gyr(i).Hits.sample)-1

        [pks,locs,w,p] =
findpeaks(gyrNorm_filt(i).trial(pole_events_gyr(i).Hits.sample(j,1)+windows(j):pole_events_gyr(i).Hits.sample(j+1,1)-
windows(j)), 'MinPeakProminence',100, 'MinPeakDistance',220);
%
figure()
%
t=(0:length(gyrNorm_filt(i).trial(pole_events_gyr(i).Hits.sample(j,1)+windows(j):pole_events_gyr(i).Hits.sample(j+1,1)-
windows(j)))-1).* (1/Fs).*1000;
%
findpeaks(gyrNorm_filt(i).trial(pole_events_gyr(i).Hits.sample(j,1)+windows(j):pole_events_gyr(i).Hits.sample(j+1,1)-
windows(j)),t,'MinPeakProminence',100*(1/Fs)*1000);
%
title ('window between two hits (filtered signal)')
%
xlabel('time (ms)')
%
ylabel('angular velocity norm (dps)')
%
if isempty(locs)==1

pole_events_gyr(i).Leaves.sample(j,1)=pole_events_gyr(i).Hits.sample(j)+round((pole_events_gyr(i).Hits.sample(j+1)-pole_events_gyr(i).Hits.sample(j)).*(2/3));
%
else
    pole_events_gyr(i).Leaves.sample(j,1)=
locs(1)+pole_events_gyr(i).Hits.sample(j,1)+windows(j)-1;
%
end

end

pole_events_gyr(i).Leaves.peak=
gyrNorm(i).trial(pole_events_gyr(i).Leaves.sample);

end

end

```

Cycle events identification: fusion method

```

% identification Hits and Leaves from IMUs
% metodo fusion:
%   -Hits from accNorm
%   -Lifts from gyrNorm

function
[pole_events]=Hits_Leaves_IMUs_fusion(gyrNorm,accNorm,gyrNorm_filt,accNorm_filt,ac-
ccNorm_filt_high,startSample,Fs)
%% questa parte commentata usarla se si passa alla funzione la norma non filtrata
ma no se si passa la norma high e low
%filtraggio high acceleration

```

```

% fCH = 40; %cutting freq
%
% [bH,aH] = butter(2,fCH/(Fs/2),'high');
% % freqz(bH,aH,Fs,Fs)
%
% %filtraggio low acceleration
% fc = 4; %cutting freq
%
% [b,a] = butter(2,fc/(Fs/2));
% % freqz(b,a,Fs,Fs)
%
% % filtraggio angular vel
% fcg = 20; %cutting freq
%
% [bg,ag] = butter(2,fcg/(Fs/2));
% % freqz(bg,ag,Fs,Fs)
% for i=1:length(accNorm)
%
% accNorm_filt(i).trial=filtfilt(b,a,accNorm(i).trial);
% accNorm_filt_high(i).trial=filtfilt(bH,aH,accNorm(i).trial);
% gyrNorm_filt(i).trial=filtfilt(bg,ag,gyrNorm(i).trial);
%
% %
% figure()
% plot(accNorm(i).trial,'LineWidth',1)
% hold on
% plot(accNorm_filt(i).trial,'LineWidth',1.5)
% title ('accNorm raw signal VS filtered signal')
% xlabel('samples')
% ylabel('acceleration norm (g)')
%
% %
% figure()
% plot(accNorm(i).trial,'LineWidth',1)
% hold on
% plot(accNorm_filt_high(i).trial,'LineWidth',1.5)
% title ('accNorm raw signal VS filtered signal')
% xlabel('samples')
% ylabel('acceleration norm (g)')
%
% end
%% voglio usare il segnale raw per identificazione lifts -- > migliori
prestazioni
% se si volesse provare a farla con il segnale filtrato basta commentare la
% riga qui sotto e andare a fare un filtraggio appropriato nella funzione
% spectral analysis
gyrNorm_filt=gyrNorm;
accNorm_filt=accNorm;
%%
% identificazione polehits
for i=1:length(accNorm)

    [pks,locs,w,p] =
findpeaks(accNorm_filt_high(i).trial(startSample(i):end), 'MinPeakHeight', 0.10*max
(accNorm_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance', 340);
%
figure()

findpeaks(accNorm_filt_high(i).trial(startSample(i):end), 'MinPeakHeight', 0.10*max
(accNorm_filt_high(i).trial(startSample(i):end)), 'MinPeakDistance', 340);

```

```

% title ('find peaks in accNorm filtered signal')
% xlabel('samples')
% ylabel('acceleration norm (g)')
pole_events(i).Hits.sample = locs + startSample(i)-1;
pole_events(i).Hits.peak= accNorm(i).trial(pole_events(i).Hits.sample);
end

%% identificazione poleleaves accNorm
for i=1:length(accNorm_filt)
    windows=round(diff(pole_events(i).Hits.sample)./10);
    for j=1:length(pole_events(i).Hits.sample)-1

        [pks,locs,w,p] =
findpeaks(accNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+windows(j):pole_events(i).Hits.sample(j+1,1)-
windows(j)), 'MinPeakHeight',0.90*max(accNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+windows(j):pole_events(i).Hits.sample(j+1,1)-
windows(j))), 'MinPeakDistance',220);
%
figure()

findpeaks(accNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+windows:pole_events(i).Hits.sample(j+1,1)-
windows), 'MinPeakHeight',0.80*max(accNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+windows:pole_events(i).Hits.sample(j+1,1)-
windows)), 'MinPeakDistance',220);
%
title ('window between two hits (filtered signal)')
xlabel('samples')
ylabel('acceleration norm (g)')
%
if length(locs) > 1
    pole_events(i).Leaves.sample(j,1)= locs(end-
1)+pole_events(i).Hits.sample(j,1)+60-1;
%
elseif isempty(locs)==1
    pole_events(i).Leaves.sample(j,1)=
pole_events(i).Hits.sample(j)+round((pole_events(i).Hits.sample(j+1)-
pole_events(i).Hits.sample(j))./3);
%
else
    pole_events(i).Leaves.sample(j,1)=
locs(end)+pole_events(i).Hits.sample(j,1)+60-1;
%
end

if isempty(locs)==1
    pole_events(i).Leaves_acc.sample(j,1)=
pole_events(i).Hits.sample(j)+round((pole_events(i).Hits.sample(j+1)-
pole_events(i).Hits.sample(j))./3);
%
else
    pole_events(i).Leaves_acc.sample(j,1)=
locs(1)+pole_events(i).Hits.sample(j,1)+windows(j)-1;
%
end

end

pole_events(i).Leaves_acc.peak=
accNorm(i).trial(pole_events(i).Leaves_acc.sample);

%
figure()
ax1=subplot(2,1,1);

```

```

%
% plot(ax1,accNorm(i).trial)
%
% hold on
%
plot(ax1,pole_events(i).Hits.sample,pole_events(i).Hits.peak,'o',pole_events(i).Leaves.sample,pole_events(i).Leaves.peak,'square' )
%
grid on
%
ax2=subplot(2,1,2);
%
plot(ax2,CoachtechaccNorm_sinc(i).speed); ylabel({'Speed (km/h)'});
%
hold on
%
plot(ax2,smoothdata(CoachtechaccNorm_sinc(i).speed,'movmean',30));
%
grid on
%
linkaxes([ax1 ax2], 'x')

end

%% identificazione poleLeaves gyr
for i=1:length(gyrNorm)

window=round(diff(pole_events(i).Hits.sample)./10);

for j=1:length(pole_events(i).Hits.sample)-1

[pks,locs,w,p] =
findpeaks(gyrNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+window(j):pole_events(i).Hits.sample(j+1,1)-
window(j)), 'MinPeakProminence',100, 'MinPeakDistance',220);
%
figure()

findpeaks(gyrNorm_filt(i).trial(pole_events(i).Hits.sample(j,1)+window(j):pole_events(i).Hits.sample(j+1,1)-
window(j)), 'MinPeakProminence',100, 'MinPeakDistance',220);
%
title ('window between two leaves (raw signal)')
%
xlabel('samples')
%
ylabel('angular velocity norm (dps)')
%
if isempty(locs)==1

pole_events(i).Leaves_gyr.sample(j,1)=pole_events(i).Hits.sample(j)+round((pole_events(i).Hits.sample(j+1)-pole_events(i).Hits.sample(j)).*(2/3));
%
else
    pole_events(i).Leaves_gyr.sample(j,1)=
locs(1)+pole_events(i).Hits.sample(j,1)+window(j)-1;
%
end
%
%versione fusion che calcola la media identificazione acc e gyr per
% i lifts. Commentare se si usa l'altra versione

pole_events(i).Leaves.sample(j,1)=round((pole_events(i).Leaves_gyr.sample(j,1)+
pole_events(i).Leaves_acc.sample(j,1))/2);
%
% versione fusion che identifica i lifts come il metodo gyr.
%
pole_events(i).Leaves.sample(j,1)=pole_events(i).Leaves_gyr.sample(j,1);
%
end
%
pole_events(i).Leaves_gyr.peak=
gyrNorm(i).trial(pole_events(i).Leaves_gyr.sample);
%
pole_events(i).Leaves.peak= gyrNorm(i).trial(pole_events(i).Leaves.sample);

end

```

```

%% pole hits refine
window1=5; % campioni di cui mi sposto dall'hit
window2=10; % finestra in campioni in cui identifico la variazione della
pendenza curva

% smooth di signal

for i=1:length(accNorm)

    accNorm_smoothed(i).trial=smooth(accNorm(i).trial,5);
    figure()
    plot(accNorm(i).trial)
    hold on
    plot(accNorm_smoothed(i).trial)
end

for i=1:length(accNorm_smoothed)

    for j=1:length(pole_events(i).Hits.sample)

        dev(i).samples(j).cycle=pole_events(i).Hits.sample(j,1)-
window2:pole_events(i).Hits.sample(j,1)-window1-1;

        dev(i).values(j).cycle=diff(accNorm_smoothed(i).trial(pole_events(i).Hits.sample(j,1)-window2:pole_events(i).Hits.sample(j,1)-window1));

        outliers=dev(i).samples(j).cycle(isoutlier(dev(i).values(j).cycle,'percentiles',[0,88]));
        pole_events_refined(i).Hits.sample(j,1)=outliers(1);
    end
    pole_events_refined(i).Leaves.sample=pole_events(i).Leaves.sample;

pole_events_refined(i).Leaves.peak=accNorm(i).trial(pole_events_refined(i).Leaves.sample);

pole_events_refined(i).Hits.peak=accNorm(i).trial(pole_events_refined(i).Hits.sample);

end
pole_events= pole_events_refined;

end

```