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# **signalanalysis**

***Release 0.1***

**Philip Gemmell**

**Oct 08, 2021**



## **CONTENTS:**



**signalanalysis** is a library including various tools for the reading, analysis and plotting of ECG and VCG data. It is designed to be as agnostic as possible for the types of data that it can read. Currently, it can read ECG data from:

1. CARP simulations of whole torso activity, using existing projects from *CARPutil* <<https://git.opencarp.org/openCARP/carputils>>;
2. .csv and .dat records;
3. wfdb file formats, using *wfdb-python* <<https://github.com/MIT-LCP/wfdb-python>>

Futher details of how to use these functions are in *Usage*, with the finer points within the files themselves.



## 1.1 Installation & Getting Started

To use `signalanalysis`, it is highly recommended to install using `pipenv`, the virtual environment, to ensure that dependencies are where possible maintained. Clone the repository, then install the requirements as follows:

```
user@home:~$ git clone git@github.com:philip-gemmell/signalanalysis.git
user@home:~$ cd signalanalysis
user@home:~/signalanalysis$ pipenv install
```

Once the repository is cloned, it is currently the case that all work must be done within the Python3 environment. However, it is recommended to use the virtual environment from `pipenv` rather than the system-wide Python3 (after entering a `pipenv` shell, it is quit using the `exit` command as shown)

```
user@home:~/signalanalysis$ pipenv shell
(signalanalysis) user@home:~/signalanalysis$ python3
>>> import signalanalysis
>>> quit()
(signalanalysis) user@home:~/signalanalysis$ exit
user@home:~/signalanalysis$
```

The project is arranged into various subdivisions. The required analysis/plotting packages for the ECG/VCG are separated out, and require separate importing. See the next section, and individual help files for further details.

- `signalanalysis`
  - `signalanalysis.general`
  - `signalanalysis.ecg`
  - `signalanalysis.vcg`
- `signalplot`
- `tools`

## 1.2 Reading ECG/VCG data

Currently, only ECG reading is supported; VCG data is calculated from ECG data using the Kors method (see method). ECG files are read upon the instantiation of the ECG class, though it is possible to re-read data if required for some reason.

```
>>> import signalanalysis.ecg
>>> import signalanalysis.vcg
>>> ecg_example = signalanalysis.ecg.Ecg("filename")
>>> vcg_example = signalanalysis.vcg.Vcg(ecg_example)
```

## 1.3 Shifting to classes from methods

Previously, all ECG/VCG data was extracted and stored in DataFrames, and most of the modules in this code currently support this format. However, it is planned to shift the main focus of the project to use classes, which allow encapsulation of linked data in one data structure. While the focus of this documentation will be future-facing, and look at using the classes, note that sometimes access to the raw, underlying DataFrames will still be required. To that end, the raw data can be accessed as the `.data` attribute:

```
>>> ecg_class = signalanalysis.ecg.Ecg("filename") # Returns an Ecg class
>>> vcg_class = signalanalysis.vcg.Vcg(ecg_class)  # Returns a Vcg class
>>> ecg_data = ecg_class.data                      # Returns a Pandas DataFrame of the
↳underlying data
>>> vcg_data = vcg_class.data                      # Returns a Pandas DataFrame of the
↳underlying data
```



## DOCUTILS DOCUMENTATION

This is the index of all modules and their associated methods and classes, documenting their use, parameters and return values. See *Usage* for a more step-by-step introduction to the intended use cases.

Broadly speaking, the modules are split thus:

- `signalanalysis` covers the analysis scripts for ECG/VCG analysis (e.g. calculating QRS duration)
- `signalplot` covers plotting methods (e.g. plotting the ECG leads on a single figure with annotation, plotting a 3D plot of VCG (including animation!))
- `tools` covers more general use tools that are not limited to ECG/VCG analysis.

*signalanalysis*

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*signalplot*

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*tools*

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## 2.1 signalanalysis

*signalanalysis.ecg*

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*signalanalysis.general*

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*signalanalysis.vcg*

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### 2.1.1 signalanalysis.ecg

#### Functions

<i>get_ecg_from_electrodes</i>	Converts electrode <code>phi_e</code> data to ECG lead data
<i>get_electrode_phi_e</i>	Extract <code>phi_e</code> data corresponding to ECG electrode locations
<i>get_qrs_start</i>	Calculates start of QRS complex using method of Hermans et al. (2017).

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Table 3 – continued from previous page

<code>read_ecg_from_csv</code>	Extract ECG data from CSV file exported from St Jude Medical ECG recording
<code>read_ecg_from_dat</code>	Read ECG data from .dat file
<code>read_ecg_from_igb</code>	Translate the phie.igb file(s) to 10-lead, 12-trace ECG data

## signalanalysis.ecg.get\_ecg\_from\_electrodes

`signalanalysis.ecg.get_ecg_from_electrodes`(*electrode\_data*: *pandas.core.frame.DataFrame*) → *pandas.core.frame.DataFrame*

Converts electrode phi\_e data to ECG lead data

Takes dictionary of phi\_e data for 10-lead ECG, and converts these data to standard ECG trace data

**Parameters** `electrode_data` (*pd.DataFrame*) – Dictionary with keys corresponding to lead locations

**Returns** `ecg` – Dictionary with keys corresponding to the ECG traces

**Return type** *pd.DataFrame*

## signalanalysis.ecg.get\_electrode\_phie

`signalanalysis.ecg.get_electrode_phie`(*phie\_data*: *numpy.ndarray*, *electrode\_file*: *Optional[str] = None*) → *pandas.core.frame.DataFrame*

Extract phi\_e data corresponding to ECG electrode locations

**Parameters**

- **phie\_data** (*np.ndarray*) – Numpy array that holds all phie data for all nodes in a given mesh
- **electrode\_file** (*str*, *optional*) – File containing entries corresponding to the nodes of the mesh which determine the location of the 10 leads for the ECG. Will default to very project specific location. The input text file has each node on a separate line (zero-indexed), with the node locations given in order: V1, V2, V3, V4, V5, V6, RA, LA, RL, LL. Will default to '12LeadElectrodes.dat', but this is almost certainly not going to right for an individual project

**Returns** `electrode_data` – Dataframe of phie data for each node, with the dictionary key labelling which node it is.

**Return type** *pd.DataFrame*

## signalanalysis.ecg.get\_qrs\_start

`signalanalysis.ecg.get_qrs_start`(*ecgs*: *Union[pandas.core.frame.DataFrame, List[pandas.core.frame.DataFrame]]*, *unipolar\_only*: *bool = True*, *plot\_result*: *bool = False*) → *List[float]*

Calculates start of QRS complex using method of Hermans et al. (2017)

Calculates the start of the QRS complex by a simplified version of the work presented in<sup>1</sup>, wherein the point of maximum second derivative of the ECG RMS signal is used as the start of the QRS complex

<sup>1</sup> Hermans BJM, Vink AS, Bennis FC, Filippini LH, Meijborg VMF, Wilde AAM, Pison L, Postema PG, Delhaas T, "The development and validation of an easy to use automatic QT-interval algorithm," PLoS ONE, 12(9), 1–14 (2017), <https://doi.org/10.1371/journal.pone.0184352>

### Parameters

- **ecgs** (*pd.DataFrame or list of pd.DataFrame*) – ECG data to analyse
- **unipolar\_only** (*bool, optional*) – Whether to use only unipolar leads to calculate RMS, default=True
- **plot\_result** (*bool, optional*) – Whether to plot the results for error-checking, default=False

**Returns** `qrs_starts` – QRS start times

**Return type** list of float

### Notes

For further details of the action of `unipolar_only`, see `general_analysis.get_signal_rms`

It is faster to use `scipy.ndimage.laplace()` rather than `np.gradient(np.gradient())`, but preliminary checks indicated some edge problems that might throw off the results.

### References

#### signalanalysis.ecg.read\_ecg\_from\_csv

`signalanalysis.ecg.read_ecg_from_csv(filename: str, normalise: bool = False) → pandas.core.frame.DataFrame`

Extract ECG data from CSV file exported from St Jude Medical ECG recording

### Parameters

- **filename** (*str*) – Name/location of the .dat file to read
- **normalise** (*bool, optional*) – Whether or not to normalise the ECG signals on a per-lead basis, default=True

**Returns** `ecg` – Extracted data for the 12-lead ECG

**Return type** list of `pd.DataFrame`

#### signalanalysis.ecg.read\_ecg\_from\_dat

`signalanalysis.ecg.read_ecg_from_dat(filename: str, normalise: bool = False) → pandas.core.frame.DataFrame`

Read ECG data from .dat file

### Parameters

- **filename** (*str*) – Name/location of the .dat file to read
- **normalise** (*bool, optional*) – Whether or not to normalise the ECG signals on a per-lead basis, default=False

**Returns** `ecg` – Extracted data for the 12-lead ECG

**Return type** `pd.DataFrame`

## signalanalysis.ecg.read\_ecg\_from\_igb

`signalanalysis.ecg.read_ecg_from_igb(filename: str, electrode_file: Optional[str] = None, normalise: bool = False, dt: float = 0.002) → pandas.core.frame.DataFrame`

Translate the phie.igb file(s) to 10-lead, 12-trace ECG data

Extracts the complete mesh data from the phie.igb file using CARPutils, which contains the data for the body surface potential for an entire human torso, before then extracting only those nodes that are relevant to the 12-lead ECG, before converting to the ECG itself <https://carpentry.medunigraz.at/carputils/generated/carputils.carpio.igb.IGBFile.html#carputils.carpio.igb.IGBFile>

### Parameters

- **filename** (*str*) – Filename for the phie.igb data to extract
- **electrode\_file** (*str*, *optional*) – File which contains the node indices in the mesh that correspond to the placement of the leads for the 10-lead ECG. Default given in `get_electrode_phie` function.
- **normalise** (*bool*, *optional*) – Whether or not to normalise the ECG signals on a per-lead basis, default=False
- **dt** (*float*, *optional*) – Time interval from which to construct the time data to associate with the ECG, default=0.002s (2ms)

**Returns** `ecgs` – DataFrame with Vm data for each of the labelled leads (the dictionary keys are the names of the leads)

**Return type** `pd.DataFrame`

## Classes

*Ecg*

Base ECG class to encapsulate data regarding an ECG recording, inheriting from `signalanalysis.general.Signal`

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## signalanalysis.ecg.Ecg

**class** `signalanalysis.ecg.Ecg(filename: str, **kwargs)`

Bases: `signalanalysis.general.Signal`

Base ECG class to encapsulate data regarding an ECG recording, inheriting from `signalanalysis.general.Signal`

### data

Raw ECG data for the different leads

**Type** `pd.DataFrame`

### filename

Filename for the location of the data

**Type** `str`

### normalised

Whether or not the data for the leads have been normalised

**Type** `bool`

### n\_beats

Number of beats recorded in the trace. Set to 0 if not calculated

**Type** int

**qrs\_start**

Times calculated for the start of the QRS complex

**Type** list of float

**qrs\_end**

Times calculated for the end of the QRS complex

**Type** end

**data\_source**

Source for the data, if known e.g. Staff III database, CARP simulation, etc.

**Type** str

**comments**

Any further details known about the data, e.g. sex, age, etc.

**Type** str

**read**(filename)

Reads in the data from the original file. Called upon initialisation

**read\_ecg\_from\_wfdb**(filename, normalise=False)

Reads data from a WFDB type series of files, e.g. from the Lobachevsky ECG database (<https://physionet.org/content/ludb/1.0.1/>)

**get\_n\_beats**(threshold=0.5, min\_separation=0.2)

Calculates the number of beats given in the recording

**get\_qrs\_start**()

Calculates the start of the QRS complex

## Methods

<i>get_n_beats</i>	Calculate the number of beats in an ECG trace, and save the individual beats to file for later use
<i>get_qrs_start</i>	Calculates start of QRS complex using method of Hermans et al. (2017).
<i>get_rms</i>	Returns the RMS of the combined signal
<i>read</i>	
<i>read_ecg_from_wfdb</i>	
<i>reset</i>	Reset all properties of the class

**get\_n\_beats**(threshold: float = 0.5, min\_separation: float = 0.2, unipolar\_only: bool = True, plot: bool = False)

Calculate the number of beats in an ECG trace, and save the individual beats to file for later use

When given the raw data of an ECG trace, will estimate the number of beats recorded in the trace based on the RMS of the ECG signal exceeding a threshold value. The estimated individual beats will then be saved in a list in a lossless manner, i.e. saved as [ECG1, ECG2, ..., ECG(n)], where ECG1=[0:peak2], ECG2=[peak1:peak3], ..., ECGn=[peak(n-1):end]

**threshold** [float {0<1}] Minimum value to search for for a peak in RMS signal to determine when a beat has occurred, default=0.5

**min\_separation** [float] Minimum time (in s) that should be used to separate separate beats, default=0.2s

**unipolar\_only** [bool, optional] Whether to use only unipolar ECG leads to calculate RMS, default=True

**plot** [bool] Whether to plot results of beat detection, default=False

**self.n\_beats** [int] Number of beats detected in signal

The scalar RMS is calculated according to

$$\text{rac}\{1\}\{n\}\text{sum}_{i=1}^n (\text{extnormal}\{\text{ECG}\}_i^2(t))$$

for all leads available from the signal (12 for ECG, 3 for VCG). If `unipolar_only` is set to true, then ECG RMS is calculated using only ‘unipolar’ leads. This uses V1-6, and the non-augmented limb leads (VF, VL and VR)

..math:: VF = LL-V\_{\{WCT\}} =

**rac{2}{3}aVF** ..math:: VL = LA-V\_{\{WCT\}} =

**rac{2}{3}aVL** ..math:: VR = RA-V\_{\{WCT\}} =

**rac{2}{3}aVR**

**get\_qrs\_start** (*unipolar\_only: bool = True, min\_separation: float = 0.05, plot\_result: bool = False*)  
Calculates start of QRS complex using method of Hermans et al. (2017)

Calculates the start of the QRS complex by a simplified version of the work presented in<sup>1</sup>, wherein the point of maximum second derivative of the ECG RMS signal is used as the start of the QRS complex

#### Parameters

- **self** (Ecg) – ECG data to analyse
- **unipolar\_only** (*bool, optional*) – Whether to use only unipolar leads to calculate RMS, default=True
- **min\_separation** (*float, optional*) – Minimum separation from the peak used to detect various beats, default=0.05s
- **plot\_result** (*bool, optional*) – Whether to plot the results for error-checking, default=False

**Returns** **self.qrs\_start** – QRS start times

**Return type** list of float

## Notes

For further details of the action of `unipolar_only`, see `general_analysis.get_signal_rms`

It is faster to use `scipy.ndimage.laplace()` rather than `np.gradient(np.gradient())`, but preliminary checks indicated some edge problems that might throw off the results.

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<sup>1</sup> Hermans BJM, Vink AS, Bennis FC, Filippini LH, Meijborg VMF, Wilde AAM, Pison L, Postema PG, Delhaas T, “The development and validation of an easy to use automatic QT-interval algorithm,” PLoS ONE, 12(9), 1–14 (2017), <https://doi.org/10.1371/journal.pone.0184352>

## References

**get\_rms**(*preprocess\_data*: *Optional[pandas.core.frame.DataFrame] = None*, *drop\_columns*: *Optional[List[str]] = None*, *unipolar\_only*: *bool = True*)

Returns the RMS of the combined signal

### Parameters

- **preprocess\_data** (*pd.DataFrame*, *optional*) – Only passed if there is some extant data that is to be used for getting the RMS (for example, if the unipolar data only from ECG is being used, and the data is thus preprocessed in a manner specific for ECG data in the ECG routine)
- **drop\_columns** (*list of str*, *optional*) – List of any columns to drop from the raw data before calculating the RMS. Can be used in conjunction with preprocess\_data
- **unipolar\_only** (*#*) –
- **RMS** (*# Whether to use only unipolar ECG leads to calculate*) –
- **default=True** –

**reset()**

Reset all properties of the class

Function called when reading in new data into an existing class (for some reason), which would make these properties and attributes clash with the other data

## 2.1.2 signalanalysis.general

### Functions

<i>get_signal_rms</i>	Calculate the ECG(RMS) of the ECG as a scalar
<i>get_twave_end</i>	Return the time point at which it is estimated that the T-wave has been completed

### signalanalysis.general.get\_signal\_rms

**signalanalysis.general.get\_signal\_rms**(*signal*: *pandas.core.frame.DataFrame*, *unipolar\_only*: *bool = True*) → *List[float]*

Calculate the ECG(RMS) of the ECG as a scalar

**signal**: *pd.DataFrame* ECG or VCG data to process

**unipolar\_only** [*bool*, *optional*] Whether to use only unipolar ECG leads to calculate RMS, default=True

**signal\_rms** [*list of float*] Scalar RMS ECG or VCG data

The scalar RMS is calculated according to

$$\frac{1}{n} \sum_{i=1}^n (\text{extnormal}\{\text{ECG}\}_i^2(t))$$

for all leads available from the signal (12 for ECG, 3 for VCG). If unipolar\_only is set to true, then ECG RMS is calculated using only ‘unipolar’ leads. This uses V1-6, and the non-augmented limb leads (VF, VL and VR)

$$\text{..math:: } VF = LL - V_{\{WCT\}} =$$

$\text{rac}\{2\}\{3\}\text{aVF} \dots \text{VL} = \text{LA} - \text{V}_{\{\text{WCT}\}} =$

$\text{rac}\{2\}\{3\}\text{aVL} \dots \text{VR} = \text{RA} - \text{V}_{\{\text{WCT}\}} =$

$\text{rac}\{2\}\{3\}\text{aVR}$

### The development and validation of an easy to use automatic QT-interval algorithm

Hermans BJM, Vink AS, Bennis FC, Filippini LH, Meijborg VMF, Wilde AAM, Pison L, Postema PG, Delhaas T PLoS ONE, 12(9), 1–14 (2017)  
<https://doi.org/10.1371/journal.pone.0184352>

## signalanalysis.general.get\_twave\_end

```
signalanalysis.general.get_twave_end(ecgs: Union[List[pandas.core.frame.DataFrame],
pandas.core.frame.DataFrame], leads: Union[str,
List[str]] = 'LII', i_distance: int = 200, filter_signal:
Optional[str] = None, baseline_adjust:
Optional[Union[float, List[float]]] = None,
return_median: bool = True, remove_outliers: bool =
True, plot_result: bool = False) →
List[pandas.core.frame.DataFrame]
```

Return the time point at which it is estimated that the T-wave has been completed

### Parameters

- **ecgs** (*pd.DataFrame or list of pd.DataFrame*) – Signal data, either ECG or VCG
- **leads** (*str, optional*) – Which lead to check for the T-wave - usually this is either 'LII' or 'V5', but can be set to a list of various leads. If set to 'global', then all T-wave values will be calculated. Will return all values unless return\_median flag is set. Default 'LII'
- **i\_distance** (*int, optional*) – Distance between peaks in the gradient, i.e. will direct that the function will only find the points of maximum gradient (representing T-wave, etc.) with a minimum distance given here (in terms of indices, rather than time). Helps prevent being overly sensitive to 'wobbles' in the ecg. Default=200
- **filter\_signal** (*{'butterworth', 'savitzky-golay'}, optional*) – Whether or not to apply a filter to the data prior to trying to find the actual T-wave gradient. Can pass either a Butterworth filter or a Savitzky-Golay filter, in which case the required kwargs for each can be provided. Default=None (no filter applied)
- **baseline\_adjust** (*float or list of float, optional*) – Point from which to calculate the adjusted baseline for calculating the T-wave, rather than using the zeroline. In line with Hermans et al., this is usually the start of the QRS complex, with the baseline calculated as the median amplitude of the 30ms before this point.
- **return\_median** (*bool, optional*) – Whether or not to return an average of the leads requested, default=True
- **remove\_outliers** (*bool, optional*) – Whether to remove T-wave end values that are greater than 1 standard deviation from the mean from the data. Only has an effect if more than one lead is provided, and return\_average is True. Default=True
- **plot\_result** (*bool, optional*) – Whether to plot the results or not, default=False

**Returns** *twave\_ends* – Time value for when T-wave is estimated to have ended.

**Return type** list of pd.DataFrame



## Notes

Calculates the end of the T-wave as the time at which the T-wave's maximum gradient tangent returns to the baseline. The baseline is either set to zero, or set to the median value of 30ms prior to the start of the QRS complex (the value of which has to be passed in the *baseline\_adjust* variable).

## References

## Classes

<i>Signal</i>	Base class for general signal, either ECG or VCG
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### signalanalysis.general.Signal

**class** signalanalysis.general.Signal(\*\*kwargs)

Bases: object

Base class for general signal, either ECG or VCG

**data**

Raw ECG data for the different leads

**Type** pd.DataFrame

**filename**

Filename for the location of the data

**Type** str

**normalised**

Whether or not the data for the leads have been normalised

**Type** bool

**n\_beats**

Number of beats recorded in the trace. Set to 0 if not calculated

**Type** int

**qrs\_start**

Times calculated for the start of the QRS complex

**Type** list of float

**qrs\_end**

Times calculated for the end of the QRS complex

**Type** end

**data\_source**

Source for the data, if known e.g. Staff III database, CARP simulation, etc.

**Type** str

**comments**

Any further details known about the data, e.g. sex, age, etc.

**Type** str

**get\_rms(unipolar\_only=True)**

Returns the RMS of the combined signal

## Methods

<code>get_n_beats</code>	Calculate the number of beats in an ECG trace, and save the individual beats to file for later use
<code>get_rms</code>	Returns the RMS of the combined signal
<code>reset</code>	Reset all properties of the class

**get\_n\_beats**(*threshold*: float = 0.5, *min\_separation*: float = 0.2, *unipolar\_only*: bool = True, *plot*: bool = False)

Calculate the number of beats in an ECG trace, and save the individual beats to file for later use

When given the raw data of an ECG trace, will estimate the number of beats recorded in the trace based on the RMS of the ECG signal exceeding a threshold value. The estimated individual beats will then be saved in a list in a lossless manner, i.e. saved as [ECG1, ECG2, ..., ECG(n)], where ECG1=[0:peak2], ECG2=[peak1:peak3], ..., ECGn=[peak(n-1):end]

**threshold** [float {0<1}] Minimum value to search for for a peak in RMS signal to determine when a beat has occurred, default=0.5

**min\_separation** [float] Minimum time (in s) that should be used to separate separate beats, default=0.2s

**unipolar\_only** [bool, optional] Whether to use only unipolar ECG leads to calculate RMS, default=True

**plot** [bool] Whether to plot results of beat detection, default=False

**self.n\_beats** [int] Number of beats detected in signal

The scalar RMS is calculated according to

$$\text{rac}\{1\}\{n\}\text{sum}_{\{i=1\}}^n (\text{extnormal}\{\text{ECG}\}_i^2(t))$$

for all leads available from the signal (12 for ECG, 3 for VCG). If *unipolar\_only* is set to true, then ECG RMS is calculated using only ‘unipolar’ leads. This uses V1-6, and the non-augmented limb leads (VF, VL and VR)

..math:: VF = LL-V\_{\{WCT\}} =

**rac{2}{3}aVF** ..math:: VL = LA-V\_{\{WCT\}} =

**rac{2}{3}aVL** ..math:: VR = RA-V\_{\{WCT\}} =

**rac{2}{3}aVR**

**get\_rms**(*preprocess\_data*: Optional[pandas.core.frame.DataFrame] = None, *drop\_columns*: Optional[List[str]] = None)

Returns the RMS of the combined signal

### Parameters

- **preprocess\_data** (pd.DataFrame, optional) – Only passed if there is some extant data that is to be used for getting the RMS (for example, if the unipolar data only from ECG is being used, and the data is thus preprocessed in a manner specific for ECG data in the ECG routine)
- **drop\_columns** (list of str, optional) – List of any columns to drop from the raw data before calculating the RMS. Can be used in conjunction with preprocess\_data
- **unipolar\_only** (#) –
- **RMS** (# Whether to use only unipolar ECG leads to calculate) –
- **default=True** –

**reset()**

Reset all properties of the class

Function called when reading in new data into an existing class (for some reason), which would make these properties and attributes clash with the other data

### 2.1.3 signalanalysis.vcg

#### Functions

<i>calculate_delta_dipole_angle</i>	Calculates the angular difference between two VCGs based on difference in azimuthal and elevation angles.
<i>compare_dipole_angles</i>	Calculates the angular differences between two VCGs at multiple points during their evolution
<i>get_azimuth_elevation</i>	Calculate azimuth and elevation angles for a specified section of the VCG.
<i>get_dipole_magnitudes</i>	Calculates metrics relating to the magnitude of the weighted dipole of the VCG
<i>get_qrs_start_end</i>	Calculate the extent of the VCG QRS complex on the basis of max derivative
<i>get_single_vcg_azimuth_elevation</i>	Get the azimuth and elevation data for a single VCG trace, along with the average dipole magnitude.
<i>get_spatial_velocity</i>	Calculate spatial velocity
<i>get_vcg_area</i>	Calculate area under VCG curve for a given section (e.g.
<i>get_vcg_from_ecg</i>	Convert ECG data to vectorcardiogram (VCG) data using the Kors matrix method
<i>get_weighted_dipole_angles</i>	Calculate metrics relating to the angles of the weighted dipole of the VCG.
<i>plot_density_effect</i>	Plot the effect of density on metrics.
<i>plot_metric_change</i>	Function to plot all the various figures for trend analysis in one go.
<i>plot_metric_change_barplot</i>	Plots a bar chart for the observed metrics.

#### signalanalysis.vcg.calculate\_delta\_dipole\_angle

signalanalysis.vcg.calculate\_delta\_dipole\_angle(*azimuth1*: List[float], *elevation1*: List[float], *azimuth2*: List[float], *elevation2*: List[float], *convert\_to\_degrees*: bool = False) → List[float]

Calculates the angular difference between two VCGs based on difference in azimuthal and elevation angles.

Useful for calculating difference between weighted averages.

#### Parameters

- **azimuth1** (*list of float*) – Azimuth angles for the first dipole
- **elevation1** (*list of float*) – Elevation angles for the first dipole
- **azimuth2** (*list of float*) – Azimuth angles for the second dipole
- **elevation2** (*list of float*) – Elevation angles for the second dipole

- **convert\_to\_degrees** (*bool, optional*) – Whether to convert the angle from radians to degrees, default=False

**Returns** **dt** – List of angles between a series of dipoles, either in radians (default) or degrees depending on input argument

**Return type** list of float

### signalanalysis.vcg.compare\_dipole\_angles

`signalanalysis.vcg.compare_dipole_angles(vcg1: pandas.core.frame.DataFrame, vcg2: pandas.core.frame.DataFrame, t_start1: float = 0, t_end1: Optional[float] = None, t_start2: float = 0, t_end2: Optional[float] = None, n_compare: int = 10, convert_to_degrees: bool = False, matlab_match: bool = False) → List[float]`

Calculates the angular differences between two VCGs at multiple points during their evolution

To compensate for the fact that the two VCG traces may not be of the same length, the comparison does not occur at every moment of the VCG; rather, the dipoles are calculated for certain fractional points during the VCG.

#### Parameters

- **vcg1** (*pd.DataFrame*) – First VCG trace to consider
- **vcg2** (*pd.DataFrame*) – Second VCG trace to consider
- **t\_start1** (*float, optional*) – Time from which to consider the data from the first VCG trace, default=0
- **t\_end1** (*float, optional*) – Time until which to consider the data from the first VCG trace, default=end
- **t\_start2** (*float, optional*) – Time from which to consider the data from the second VCG trace, default=0
- **t\_end2** (*float, optional*) – Time until which to consider the data from the second VCG trace, default=end
- **n\_compare** (*int, optional*) – Number of points during the VCGs at which to calculate the dipole angle. If set to -1, will calculate at every point during the VCG, but requires VCG traces to be the same length, default=10
- **convert\_to\_degrees** (*bool, optional*) – Whether to convert the angles from radians to degrees, default=False
- **matlab\_match** (*bool, optional*) – Whether to extract the data segment to match Matlab output or to use simpler Python, default=False

**Returns** **dt** – Angle between two given VCGs at n points during the VCG, where n is given as input

**Return type** list of float

## signalanalysis.vcg.get\_azimuth\_elevation

```
signalanalysis.vcg.get_azimuth_elevation(vcgs:
                                         Union[List[pandas.core.frame.DataFrame],
                                         pandas.core.frame.DataFrame], t_start:
                                         Optional[List[float]] = None, t_end:
                                         Optional[List[float]] = None) →
                                         Tuple[List[Iterable[float]], List[Iterable[float]]]
```

Calculate azimuth and elevation angles for a specified section of the VCG.

Will calculate the azimuth and elevation angles for the VCG at each recorded point, potentially within specified limits (e.g. start/end of QRS)

### Parameters

- **vcgs** (*pd.DataFrame or list of pd.DataFrame*) – VCG data to calculate
- **t\_start** (*list of float, optional*) – Start time from which to calculate the angles, default=0
- **t\_end** (*list of float, optional*) – End time until which to calculate the angles, default=end

### Returns

- **azimuth** (*list of list of float*) – List (one entry for each passed VCG) of azimuth angles (in radians) for the dipole for every time point during the specified range
- **elevation** (*list of list of float*) – List (one entry for each passed VCG) of elevation angles (in radians) for the dipole for every time point during the specified range

## signalanalysis.vcg.get\_dipole\_magnitudes

```
signalanalysis.vcg.get_dipole_magnitudes(vcgs:
                                         Union[List[pandas.core.frame.DataFrame],
                                         pandas.core.frame.DataFrame], t_start:
                                         Union[float, List[float]] = 0, t_end:
                                         Union[float, List[float]] = -1) →
                                         Tuple[List[numpy.ndarray], List[float],
                                         List[float], List[List[float]], List]
```

Calculates metrics relating to the magnitude of the weighted dipole of the VCG

Returns the mean weighted dipole, maximum dipole magnitude,(x,y,z) components of the maximum dipole and the time at which the maximum dipole occurs

### Parameters

- **vcgs** (*pd.DataFrame or list of pd.DataFrame*) – VCG data to calculate
- **t\_start** (*list of float, optional*) – Start time from which to calculate the magnitude, default=0 (for any other value to be recognisable, time variable must be given)
- **t\_end** (*list of float, optional*) – End time until which to calculate the magnitudes, default=end (for any other value to be recognisable, time variable must be given)

### Returns

- **dipole\_magnitude** (*list of np.ndarray*) – Magnitude time courses for each VCG

- **weighted\_magnitude** (*list of float*) – Mean magnitude of the VCG
- **max\_dipole\_magnitude** (*list of float*) – Maximum magnitude of the VCG
- **max\_dipole\_components** (*list of list of float*) – x, y, z components of the dipole at is maximum value
- **max\_dipole\_time** (*list of float*) – Time at which the maximum magnitude of the VCG occurs

### signalanalysis.vcg.get\_qrs\_start\_end

```
signalanalysis.vcg.get_qrs_start_end(vcgs: Union[List[pandas.core.frame.DataFrame],
                                                pandas.core.frame.DataFrame], velocity_offset: int
                                     = 2, low_p: float = 40, order: int = 2,
                                     threshold_frac_start: float = 0.22,
                                     threshold_frac_end: float = 0.54, filter_sv: bool =
                                     True, qrs_window: float = 180, ecgs: Op-
                                     tional[Union[List[pandas.core.frame.DataFrame],
                                                pandas.core.frame.DataFrame]] = None) →
                                     Tuple[List[float], List[float], List[float]]
```

Calculate the extent of the VCG QRS complex on the basis of max derivative

TODO: Check whether i\_qrs\_start variable is needed, or can be simplified using DataFrame function

Calculate the start and end points, and hence duration, of the QRS complex of a list of VCGs. It does this by finding the time at which the spatial velocity of the VCG exceeds a threshold value (the start time), then searches backwards from the end of the VCG to find when this threshold is exceeded (the end time); the start and end thresholds do not necessarily have to be the same.

#### Parameters

- **vcgs** (*list of pd.DataFrame or pd.DataFrame*) – List of VCG data to get QRS start and end points for
- **velocity\_offset** (*int, optional*) – Offset between values in VCG over which to calculate spatial velocity, i.e. 1 will use neighbouring values to calculate the gradient/velocity. Default=2
- **low\_p** (*float, optional*) – Low frequency for bandpass filter, default=40
- **order** (*int, optional*) – Order for Butterworth filter, default=2
- **threshold\_frac\_start** (*float, optional*) – Fraction of maximum spatial velocity to trigger start of QRS detection, default=0.15
- **threshold\_frac\_end** (*float, optional*) – Fraction of maximum spatial velocity to trigger end of QRS detection, default=0.15
- **filter\_sv** (*bool, optional*) – Whether or not to apply filtering to spatial velocity prior to finding the start/end points for the threshold
- **qrs\_window** (*float, optional*) – Default size of ‘window’ in which to search for end of QRS complex, default=180ms
- **ecgs** (*list of pd.DataFrame or pd.DataFrame, optional*) – ECG data associated with VCG data. Only used if having trouble establishing QRS start, in which case will be used to plot ECG data to allow user to determine whether or not the QRS is occurring at the start of the simulation, or whether there is a more deep-seated issue with the data.

### Returns

- **qrs\_start** (*list of float*) – List of start time of QRS complexes of provided VCGs
- **qrs\_end** (*list of float*) – List of end time of QRS complex of provided VCGs
- **qrs\_duration** (*list of float*) – List of duration of QRS complex of provided VCGs

## signalanalysis.vcg.get\_single\_vcg\_azimuth\_elevation

`signalanalysis.vcg.get_single_vcg_azimuth_elevation(vcg:`  
*pandas.core.frame.DataFrame,*  
*t\_start: float, t\_end: float,*  
*weighted: bool = True) →*  
*Tuple[List[float], List[float],*  
*numpy.ndarray]*

Get the azimuth and elevation data for a single VCG trace, along with the average dipole magnitude.

Returns the azimuth and elevation angles for a single given VCG trace. Can analyse only a segment of the VCG if required, and can weight the angles according to the dipole magnitude. Primarily designed as a helper function for `get_azimuth_elevation` and `get_weighted_dipole_angles`.

### Parameters

- **vcg** (*pd.DataFrame*) – VCG data to calculate
- **t\_start** (*float*) – Start time from which to calculate the angles
- **t\_end** (*float*) – End time until which to calculate the angles
- **weighted** (*bool, optional*) – Whether or not to weight the returned angles by the magnitude of the dipole at the same moment, default=True

### Returns

- **theta** (*list of float*) – List of the azimuth angles for the VCG dipole, potentially weighted according to the dipole magnitude at the associated time
- **phi** (*list of float*) – List of the elevation above xy-plane angles for the VCG dipole, potentially weighted according to the dipole magnitude at the associated time
- **dipole\_magnitude** (*np.ndarray*) – Array containing the dipole magnitude at all points throughout the VCG

## signalanalysis.vcg.get\_spatial\_velocity

`signalanalysis.vcg.get_spatial_velocity(vcgs:`  
*Union[List[pandas.core.frame.DataFrame],*  
*pandas.core.frame.DataFrame], velocity\_offset:*  
*int = 2, filter\_sv: bool = True, low\_p: float =*  
*40, order: int = 2) →*  
*List[pandas.core.frame.DataFrame]*

Calculate spatial velocity

Calculate the spatial velocity of a VCG, in terms of calculating the gradient of the VCG in each of its x, y and z components, before combining these components in a Euclidian norm. Will then find the point at which the spatial velocity exceeds a threshold value, and the point at which it declines below another threshold value.

### Parameters

- **vcgs** (*list of pd.DataFrame or pd.DataFrame*) – VCG data to analyse
- **velocity\_offset** (*int, optional*) – Offset between values in VCG over which to calculate spatial velocity, i.e. 1 will use neighbouring values to calculate the gradient/velocity. Default=2
- **filter\_sv** (*bool, optional*) – Whether or not to apply filtering to spatial velocity, default=True
- **low\_p** (*float, optional*) – Low frequency for bandpass filter, default=40
- **order** (*int, optional*) – Order for Butterworth filter, default=2

**Returns** *sv* – Spatial velocity data, filtered according to input parameters

**Return type** list of pd.DataFrame

### Notes

Calculation of spatial velocity based on<sup>1,2,3</sup>

### References

#### signalanalysis.vcg.get\_vcg\_area

```
signalanalysis.vcg.get_vcg_area(vcgs: Union[List[pandas.core.frame.DataFrame],
                                             pandas.core.frame.DataFrame], limits_start:
                                Optional[List[float]] = None, limits_end:
                                Optional[List[float]] = None, method: str = 'pythag',
                                matlab_match: bool = False) → List[float]
```

Calculate area under VCG curve for a given section (e.g. QRS complex).

Calculate the area under the VCG between two intervals (usually QRS start and QRS end). This is calculated in two ways: a 'Pythagorean' method, wherein the area under each of the VCG(x), VCG(y) and VCG(z) curves are calculated, then combined in a Euclidean norm, or a '3D' method, wherein the area of the arc traced in 3D space between successive timepoints is calculated, then summed.

### Parameters

- **vcgs** (*pd.DataFrame or list of pd.DataFrame*) – VCG data from which to get area
- **limits\_start** (*list of float, optional*) – Start times (NOT INDICES) for where to calculate area under curve from, default=0
- **limits\_end** (*list of float, optional*) – End times (NOT INDICES) for where to calculate area under curve until, default=end
- **method** (*{'pythag', '3d'}, optional*) – Which method to use to calculate the area under the VCG curve, default='pythag'

<sup>1</sup> Kors JA, van Herpen G, "Methodology of QT-interval measurement in the modular ECG analysis system (MEANS)" Ann Noninvasive Electrocardiol. 2009 Jan;14 Suppl 1:S48-53. doi: 10.1111/j.1542-474X.2008.00261.x.

<sup>2</sup> Xue JQ, "Robust QT Interval Estimation—From Algorithm to Validation" Ann Noninvasive Electrocardiol. 2009 Jan;14 Suppl 1:S35-41. doi: 10.1111/j.1542-474X.2008.00264.x.

<sup>3</sup> Sörnmo L, "A model-based approach to QRS delineation" Comput Biomed Res. 1987 Dec;20(6):526-42.



- **matlab\_match** (*bool, optional*) – Whether to alter the calculation for start and end indices to match the original Matlab output, from which this module is based, default=False

#### Returns

- **qrs\_area\_3d** (*list of float*) – Values for the area under the curve (as defined by the 3D method) between the provided limits for each of the VCGs
- **qrs\_area\_pythag** (*list of float*) – Values for the area under the curve (as defined by the Pythagorean method) between the provided limits for each of the VCGs
- **qrs\_area\_components** (*list of list of float*) – Areas under the individual x, y, z curves of the VCG, for each of the supplied VCGs

### signalanalysis.vcg.get\_vcg\_from\_ecg

`signalanalysis.vcg.get_vcg_from_ecg(ecgs: Union[List[pandas.core.frame.DataFrame], pandas.core.frame.DataFrame]) → List[pandas.core.frame.DataFrame]`

Convert ECG data to vectorcardiogram (VCG) data using the Kors matrix method

**Parameters** *ecgs* (*list of pd.DataFrame or pd.DataFrame*) – List of ECG dataframe data, or ECG dataframe data directly, with dict keys corresponding to ECG outputs

**Returns** *vcgs* – List of VCG output data

**Return type** list of pd.DataFrame

#### References

**Kors JA, van Herpen G, Sittig AC, van Bommel JH.** Reconstruction of the Frank vectorcardiogram from standard electrocardiographic leads: diagnostic comparison of different methods Eur Heart J. 1990 Dec;11(12):1083-92.

### signalanalysis.vcg.get\_weighted\_dipole\_angles

`signalanalysis.vcg.get_weighted_dipole_angles(vcgs: Union[List[pandas.core.frame.DataFrame], pandas.core.frame.DataFrame], t_start: Optional[List[float]] = None, t_end: Optional[List[float]] = None) → Tuple[List[float], List[float], List[List[float]]]`

Calculate metrics relating to the angles of the weighted dipole of the VCG. Usually used with QRS limits.

Calculates the weighted averages of both the azimuth and the elevation (inclination above the xy-plane) for a given section of the VCG. Based on these weighted averages of the angles, the unit weighted dipole for that section of the VCG is returned as well.

#### Parameters

- **vcgs** (*pd.DataFrame or list of pd.DataFrame*) – VCG data to calculate

- **t\_start** (*list of float, optional*) – Start time from which to calculate the angles, default=0
- **t\_end** (*list of float, optional*) – End time until which to calculate the angles, default=end

#### Returns

- **waa** (*list of float*) – List of Weighted Average Azimuth angles (in radians) for each given VCG
- **wae** (*list of float*) – List of Weighted Average Elevation (above xy-plane) angles (in radians) for each given VCG
- **uwd** (*list of list of float*) – x, y, z coordinates for the unit mean weighted dipole for the given (section of) VCGs

### signalanalysis.vcg.plot\_density\_effect

`signalanalysis.vcg.plot_density_effect(metrics, metric_name, metric_labels=None, density_labels=None, linestyle=None, colours=None, markers=None)`

Plot the effect of density on metrics.

### signalanalysis.vcg.plot\_metric\_change

`signalanalysis.vcg.plot_metric_change(metrics, metrics_phi, metrics_rho, metrics_z, metric_name, metrics_lv=None, labels=None, scattermarkers=None, linemarkers=None, colours=None, linestyle=None, layout=None, axis_match=True, no_labels=False)`

Function to plot all the various figures for trend analysis in one go.

### signalanalysis.vcg.plot\_metric\_change\_barplot

`signalanalysis.vcg.plot_metric_change_barplot(metrics_cont, metrics_lv, metrics_sept, metric_labels, layout=None)`

Plots a bar chart for the observed metrics.

## Classes

---

*Vcg*

Base class to encapsulate data from VCG

---

## signalanalysis.vcg.Vcg

**class** signalanalysis.vcg.Vcg(*ecg*: signalanalysis.ecg.Ecg, *\*\*kwargs*)

Bases: *signalanalysis.general.Signal*

Base class to encapsulate data from VCG

## Methods

<i>get_from_ecg</i>	Convert ECG data to vectorcardiogram (VCG) data using the Kors matrix method
<i>get_n_beats</i>	Calculate the number of beats in an ECG trace, and save the individual beats to file for later use
<i>get_rms</i>	Returns the RMS of the combined signal
<i>reset</i>	Reset all properties of the class

**get\_from\_ecg**(*ecg*: signalanalysis.ecg.Ecg)

Convert ECG data to vectorcardiogram (VCG) data using the Kors matrix method

**Parameters** **ecg** (signalanalysis.ecg.Ecg) – List of ECG dataframe data, or ECG dataframe data directly, with dict keys corresponding to ECG outputs

## References

**Kors JA, van Herpen G, Sittig AC, van Bommel JH.** Reconstruction of the Frank vectorcardiogram from standard electrocardiographic leads: diagnostic comparison of different methods Eur Heart J. 1990 Dec;11(12):1083-92.

**get\_n\_beats**(*threshold*: float = 0.5, *min\_separation*: float = 0.2, *unipolar\_only*: bool = True, *plot*: bool = False)

Calculate the number of beats in an ECG trace, and save the individual beats to file for later use

When given the raw data of an ECG trace, will estimate the number of beats recorded in the trace based on the RMS of the ECG signal exceeding a threshold value. The estimated individual beats will then be saved in a list in a lossless manner, i.e. saved as [ECG1, ECG2, ..., ECG(n)], where ECG1=[0:peak2], ECG2=[peak1:peak3], ..., ECGn=[peak(n-1):end]

**threshold** [float {0<1}] Minimum value to search for for a peak in RMS signal to determine when a beat has occurred, default=0.5

**min\_separation** [float] Minimum time (in s) that should be used to separate separate beats, default=0.2s

**unipolar\_only** [bool, optional] Whether to use only unipolar ECG leads to calculate RMS, default=True

**plot** [bool] Whether to plot results of beat detection, default=False

**self.n\_beats** [int] Number of beats detected in signal

The scalar RMS is calculated according to

$$\text{rac}\{1\}\{n\}\text{sum}_{\{i=1\}}^n(\text{extnormal}\{\text{ECG}\}_i^2(t))$$

for all leads available from the signal (12 for ECG, 3 for VCG). If `unipolar_only` is set to true, then ECG RMS is calculated using only ‘unipolar’ leads. This uses V1-6, and the non-augmented limb leads (VF, VL and VR)

$$\text{..math:: } VF = LL - V_{\{WCT\}} =$$

$$\text{rac}\{2\}\{3\}\text{aVF} \text{ ..math:: } VL = LA - V_{\{WCT\}} =$$

$$\text{rac}\{2\}\{3\}\text{aVL} \text{ ..math:: } VR = RA - V_{\{WCT\}} =$$

$$\text{rac}\{2\}\{3\}\text{aVR}$$

**get\_rms**(*preprocess\_data*: *Optional[pandas.core.frame.DataFrame] = None*,  
*drop\_columns*: *Optional[List[str]] = None*)

Returns the RMS of the combined signal

#### Parameters

- **preprocess\_data** (*pd.DataFrame*, *optional*) – Only passed if there is some extant data that is to be used for getting the RMS (for example, if the unipolar data only from ECG is being used, and the data is thus preprocessed in a manner specific for ECG data in the ECG routine)
- **drop\_columns** (*list of str*, *optional*) – List of any columns to drop from the raw data before calculating the RMS. Can be used in conjunction with `preprocess_data`
- **unipolar\_only** (#) –
- **RMS** (# *Whether to use only unipolar ECG leads to calculate*) –
- **default=True** –

**reset()**

Reset all properties of the class

Function called when reading in new data into an existing class (for some reason), which would make these properties and attributes clash with the other data

## 2.2 signalplot

*signalplot.ecg*

---

*signalplot.general*

---

*signalplot.vcg*

---

### 2.2.1 signalplot.ecg

#### Functions

*plot*

Plot and label the ECG data from simulation(s).

---

## signalplot.ecg.plot

```
signalplot.ecg.plot(ecgs: Union[List[pandas.core.frame.DataFrame],
                                pandas.core.frame.DataFrame], legend_ecg: Optional[List[str]] =
                                None, linewidths_ecg: float = 2, limits: Optional[Union[list, float]]
                                = None, legend_limits: Optional[List[str]] = None, plot_sequence:
                                Optional[List[str]] = None, single_fig: bool = True, colours_ecg:
                                Optional[Union[List[str], List[List[float]], List[Tuple[float]]]] =
                                None, linestyle_ecg: Optional[List[str]] = '-', colours_limits:
                                Optional[Union[List[str], List[List[float]], List[Tuple[float]]]] =
                                None, linestyle_limits: Optional[List[str]] = None, fig:
                                Optional[matplotlib.pyplot.figure] = None, ax=None) → tuple
```

Plot and label the ECG data from simulation(s). Optional to add in QRS start/end boundaries for plotting

### Parameters

- **ecgs** (*pd.DataFrame or list of pd.DataFrame*) – Dataframe or list of dataframes for ECG data, with keys corresponding to the trace name and index to the time data
- **legend\_ecg** (*list of str, optional*) – List of names for each given set of ECG data e.g. ['BCL=300ms', 'BCL=600ms'], default=None
- **linewidths\_ecg** (*float, optional*) – Width to use for plotting lines, default=3
- **limits** (*float or list of float or pd.DataFrame, optional*) – Optional temporal limits (e.g. QRS limits) to add to ECG plots. Can add multiple limits, which will be plotted identically on all axes. If provided as a dataframe, will plot the limits on the relevant axis
- **legend\_limits** (*list of str, optional*) – List of names for each given set of limits e.g. ['QRS start', 'QRS end'], default=None
- **plot\_sequence** (*list of str, optional*) – Sequence in which to plot the ECG traces. Will default to: V1, V2, V3, V4, V5, V6, LI, LII, LIII, aVR, aVL, aVF
- **single\_fig** (*bool, optional*) – If true, will plot all axes on a single figure window. If false, will plot each axis on a separate figure window. Default is True
- **colours\_ecg** (*str or list of str or list of list/tuple of float, optional*) – Colours to be used to plot ECG traces. Can provide as either string (e.g. 'b') or as RGB values (floats). Will default to `ca.get_plot_colours()`
- **linestyles\_ecg** (*str or list, optional*) – Linestyles to be used to plot ECG traces. Will default to `ca.get_plot_lines()`
- **colours\_limits** (*str or list of str or list of list/tuple of float, optional*) – Colours to be used to plot limits. Can provide as either string (e.g. 'b') or as RGB values (floats). Will default to `ca.get_plot_colours()`
- **linestyles\_limits** (*str or list, optional*) – Linestyles to be used to plot limits. Will default to `ca.get_plot_lines()`
- **fig** (*optional*) – If given, will plot data on existing figure window

- **ax** (*optional*) – If given, will plot data using existing axis handles

#### Returns

- *fig* – Handle to output figure window, or dictionary to several handles if traces are all plotted in separate figure windows (if *single\_fig=False*)
- **ax** (*dict*) – Dictionary to axis handles for ECG traces

#### Raises

- **AssertionError** – Checks that various list lengths are the same
- **TypeError** – If input argument is given in an unexpected format

## 2.2.2 signalplot.general

## 2.2.3 signalplot.vcg

### Functions

<i>add_unit_sphere</i>	Add a unit sphere to a 3D plot
<i>animate_3d</i>	Animate the evolution of the VCG in 3D space, saving that animation to a file.
<i>plot_2d</i>	Plot x vs y (or y vs z, or other combination) for VCG trace, with line colour shifting to show time progression.
<i>plot_3d</i>	Plot the evolution of VCG in 3D space
<i>plot_arc3d</i>	Plot arc between two given vectors in 3D space.
<i>plot_density_effect</i>	Plot the effect of density on metrics.
<i>plot_metric_change</i>	Function to plot all the various figures for trend analysis in one go.
<i>plot_metric_change_barplot</i>	Plots a bar chart for the observed metrics.
<i>plot_spatial_velocity</i>	Plot the spatial velocity for given VCG data
<i>plot_xyz_components</i>	Plot x, y, z components of VCG data
<i>plot_xyz_vector</i>	Plots a specific vector in 3D space (e.g.

### signalplot.vcg.add\_unit\_sphere

signalplot.vcg.add\_unit\_sphere(ax) → None

Add a unit sphere to a 3D plot

**Parameters** **ax** – Handles to axes

## signalplot.vcg.animate\_3d

`signalplot.vcg.animate_3d(vcg: numpy.ndarray, limits: Optional[Union[float, List[float], List[List[float]]]] = None, linestyle: Optional[str] = '-', colourmap: Optional[str] = 'viridis', linewidth: Optional[float] = 3, output_file: Optional[str] = 'vcg_xyz.mp4') → None`

Animate the evolution of the VCG in 3D space, saving that animation to a file.

### Parameters

- **vcg** (*np.ndarray*) – VCG data
- **limits** (*float or list of float or list of list of floats, optional*) –

**Limits for the axes. If none, will set to the min/max values of the provided data. Can provide:**

- 1) a single value (+/- of that value applied to all axes)
- 2) [min, max] to be applied to all axes
- 3) [[xmin, xmax], [ymin, ymax], [zmin, zmax]]

- **linestyle** (*str, optional*) – Linestyle for the data, default='-'
- **colourmap** (*str, optional*) – Colourmap to use when plotting, default='viridis'
- **linewidth** (*float, optional*) – Linewidth when used to plot VCG, default=3
- **output\_file** (*str, optional*) – Name of the file to save the animation to, default='vcg\_xyz.mp4'

## signalplot.vcg.plot\_2d

`signalplot.vcg.plot_2d(vcg: pandas.core.frame.DataFrame, x_plot: str = 'x', y_plot: str = 'y', linestyle: str = '-', colourmap: str = 'viridis', linewidth: float = 3, axis_limits: Optional[Union[float, List[float]]] = None, fig: Optional[matplotlib.pyplot.figure] = None) → matplotlib.pyplot.figure`

Plot x vs y (or y vs z, or other combination) for VCG trace, with line colour shifting to show time progression.

Plot a colour-varying course of a VCG in 2D space

### Parameters

- **vcg** (*pd.DataFrame*) – VCG data to be plotted
- **x\_plot** (*str, optional*) – Which components of VCG to plot, default='x', 'y'
- **y\_plot** (*str, optional*) – Which components of VCG to plot, default='x', 'y'
- **linestyle** (*str, optional*) – Linestyle to apply to the plot, default='-'

- **colourmap** (*str, optional*) – Colourmap to use for the line, default='viridis'
- **linewidth** (*float, optional*) – Linewidth to use, default=3
- **axis\_limits** (*list of float or float, optional*) – Limits to apply to the axes, default=None
- **fig** (*plt.figure, optional*) – Handle to pre-existing figure (if present) on which to plot data, default=None

**Returns** **fig** – Handle to output figure window

**Return type** plt.figure

### signalplot.vcg.plot\_3d

```
signalplot.vcg.plot_3d(vcg: pandas.core.frame.DataFrame, linestyle: str = '-',
                       colourmap: str = 'viridis', linewidth: float = 3.0, axis_limits:
                       Optional[Union[float, List[float]]] = None, unit_min: bool =
                       True, sig_fig: Optional[int] = None, fig:
                       Optional[matplotlib.pyplot.figure] = None) →
                       matplotlib.pyplot.figure
```

Plot the evolution of VCG in 3D space

#### Parameters

- **vcg** (*pd.DataFrame*) – VCG data
- **linestyle** (*str, optional*) – Linestyle to plot data, default='-'
- **colourmap** (*str, optional*) – Colourmap to use when plotting data, default='viridis'
- **linewidth** (*float, optional*) – Linewidth to use, default=3
- **axis\_limits** (*list of float or float, optional*) – Limits to apply to the axes, default=None
- **unit\_min** (*bool, optional*) – Whether to have the axes set to, as a minimum, unit length, default=True
- **sig\_fig** (*int, optional*) – Maximum number of decimal places to be used on the axis plots (e.g., if set to 2, 0.12345 will be displayed as 0.12). Used to avoid floating point errors, default=None (no adaption made)
- **fig** (*plt.figure, optional*) – Handle to existing figure (if exists)

**Returns** **fig** – Figure handle

**Return type** plt.figure



### signalplot.vcg.plot\_arc3d

`signalplot.vcg.plot_arc3d(vector1: List[float], vector2: List[float], radius: float = 0.2, fig: Optional[matplotlib.pyplot.figure] = None, colour: str = 'C0') → matplotlib.pyplot.figure`

Plot arc between two given vectors in 3D space.

#### Parameters

- **vector1** (*list of float*) – First vector
- **vector2** (*list of float*) – Second vector
- **radius** (*float, optional*) – Radius of arc to plot on figure
- **fig** (*plt.figure, optional*) – Handle of figure on which to plot the arc. If not given, will produce new figure
- **colour** (*str, optional*) – Colour in which to display the arc

**Returns** **fig** – Handle for figure on which arc has been plotted

**Return type** `plt.figure`

### signalplot.vcg.plot\_density\_effect

`signalplot.vcg.plot_density_effect(metrics: List[List[float]], metric_name: str, metric_labels: Optional[List[str]] = None, density_labels: Optional[List[str]] = None, linestyle: Optional[List[str]] = None, colours: Optional[List[str]] = None, markers: Optional[List[str]] = None)`

Plot the effect of density on metrics.

TODO: look into decorator for the LaTeX preamble?

#### Parameters

- **metrics** (*list of list of float*) – Effects of scar density on given metrics, presented as e.g. [metric\_LV, metric\_septum]
- **metric\_name** (*str*) – Name of metric being assessed
- **metric\_labels** (*list of str, optional*) – Labels for the metrics being plotted, default=['LV', 'Septum']
- **density\_labels** (*list of str, optional*) – Labels for the different scar densities being plotted
- **linestyle** (*list of str, optional*) – Linestyles for the density effect plots, default=['-' for \_ in range(len(metrics))]
- **colours** (*list of str, optional*) – Colours to use for the plot, default=`common_analysis.get_plot_colours(len(metrics))`
- **markers** (*list of str, optional*) – Markers to use for the discrete data points in the plot, default=['o' for \_ in range(len(metrics))]

## signalplot.vcg.plot\_metric\_change

```
signalplot.vcg.plot_metric_change(metrics: List[List[List[float]]], metrics_phi:
    List[List[List[float]]], metrics_rho:
    List[List[List[float]]], metrics_z:
    List[List[List[float]]], metric_name: str,
    metrics_lv: Optional[List[bool]] = None, labels:
    Optional[List[str]] = None, scattermarkers:
    Optional[List[str]] = None, linemarkers:
    Optional[List[str]] = None, colours:
    Optional[List[str]] = None, linestyle:
    Optional[List[str]] = None, layout: Optional[str]
    = None, axis_match: bool = True, no_labels: bool
    = False) → Tuple
```

Function to plot all the various figures for trend analysis in one go.

TODO: labels parameter seems redundant - potentially remove

### Parameters

- **metrics** (*list of list of list of float*) – Complete list of all metric data recorded [phi+rho+z+size+other]
- **metrics\_phi** (*list of list of list of float*) – Metric data recorded for scar size variations in phi UVC
- **metrics\_rho** (*list of list of list of float*) – Metric data recorded for scar size variations in rho UVC
- **metrics\_z** (*list of list of list of float*) – Metric data recorded for scar size variations in z UVC
- **metric\_name** (*str*) – Name of metric being plotted (for labelling purposes). Can incorporate LaTeX typesetting.
- **metrics\_lv** (*list of bool, optional*) – Boolean to distinguish whether metrics being plotted are for LV or septal data, default=[True, False]
- **labels** (*list of str, optional*) – Labels for the data sets being plotted, default=['LV', 'Septum']
- **scattermarkers** (*list of str, optional*) – Markers to use to plot the data on the scatterplots, default=['+', 'o', 'D', 'v', '^', 's', '\*', 'x']
- **linemarkers** (*list of str, optional*) – Markers to use on the line plots to indicate discrete data points, required to be at least as long as the longest line plot to be drawn (rho), default=['.' for \_ in range(len(metrics\_rho))]
- **colours** (*list of str, optional*) – Sequence of colours to plot data (if plotting LV and septal data, will require two different colours to allow them to be distinguished), default=common\_analysis.get\_plot\_colours(len(metrics\_rho))
- **linestyles** (*list of str, optional*) – Linestyles to be used for plotting the data on lineplots, default=['-' for \_ in range(len(metrics\_rho))]

- **layout** (*{'combined', 'figures'}, optional*) – String specifying the output, whether all plots should be combined into one figure window (default), or whether individual figure windows should be plotted for each plot
- **axis\_match** (*bool, optional*) – Whether to make sure all plotted figures share the same axis ranges, default=True
- **no\_labels** (*bool, optional*) – Whether to have labels on the figures, or not - having no labels can make it far easier to ‘prettify’ the figures manually later in Inkscape, default=False

#### Returns

- **fig** (*plt.figure or dict of plt.figure*) – Handle to figure(s)
- **ax** (*dict*) – Handles to axes

### signalplot.vcg.plot\_metric\_change\_barplot

signalplot.vcg.plot\_metric\_change\_barplot(*metrics\_cont: List[List[float]],  
metrics\_lv: List[List[float]],  
metrics\_sept: List[List[float]],  
metric\_labels: List[str], layout:  
Optional[str] = None*) → Tuple

Plots a bar chart for the observed metrics.

#### Parameters

- **metrics\_cont** (*list of list of float*) – Values of series of metrics for no scar
- **metrics\_lv** (*list of list of float*) – Values of series of metrics for LV scar
- **metrics\_sept** (*list of list of float*) – Values of series of metrics for septal scar
- **metric\_labels** (*list of str*) – Names of metrics being plotted
- **layout** (*{'combined', 'fig'}, optional*) – Whether to plot bar charts on combined plot window, or in individual figure windows

#### Returns

- **fig** (*plt.figure or list of plt.figure*) – Handle(s) to figures
- **ax** (*list*) – Handles to axes

## signalplot.vcg.plot\_spatial\_velocity

```
signalplot.vcg.plot_spatial_velocity(vcg: Union[pandas.core.frame.DataFrame,
List[pandas.core.frame.DataFrame]], sv:
Optional[List[List[float]]] = None, limits:
Optional[List[List[float]]] = None, fig:
Optional[matplotlib.pyplot.figure] = None,
legend_vcg: Optional[Union[List[str], str]] =
None, legend_limits:
Optional[Union[List[str], str]] = None,
limits_linestyles: Optional[List[str]] = None,
limits_colours: Optional[List[str]] = None,
filter_sv: bool = True) → Tuple
```

Plot the spatial velocity for given VCG data

Plot the spatial velocity and VCG elements, with limits (e.g. QRS limits) if provided. Note that if spatial velocity is not provided, default values will be used to calculate it - if anything else is desired, then spatial velocity must be calculated first and provided to the function.

### Parameters

- **vcg** (*pd.DataFrame* or *list of pd.DataFrame*) – VCG data
- **sv** (*list of list of float, optional*) – Spatial velocity data. Only required to be given here if special parameters wish to be given, otherwise it will be calculated using default parameters (default)
- **limits** (*list of list of float, optional*) – A series of ‘limits’ to be plotted on the figure with the VCG and spatial plot. Presented as a list of the same length of the VCG data, with the required limits within:

e.g. [[QRS\_start1, QRS\_start2, ...], [QRS\_end1, QRS\_end2, ...], ...]

Default=None

- **fig** (*plt.figure, optional*) – Handle to existing figure, if data is wished to be plotted on existing plot, default=None
- **legend\_vcg** (*str or list of str, optional*) – Labels to apply to the VCG/SV data, default=None
- **legend\_limits** (*str or list of str, optional*) – Labels to apply to the limits, default=None
- **limits\_linestyles** (*list of str, optional*) – Linestyles to apply to the different limits being supplied, default=None (will use varying linestyles based on tools.plotting.get\_plot\_lines)
- **limits\_colours** (*list of str, optional*) – Colours to apply to the different limits being supplied, default=None (will use varying colours based on tools.plotting.get\_plot\_colours)
- **filter\_sv** (*bool, optional*) – Whether or not to apply filtering to spatial velocity prior to finding the start/end points for the threshold, default=True

**Returns** Handles to the figure and axes generated

**Return type** fig, ax

## signalplot.vcg.plot\_xyz\_components

```
signalplot.vcg.plot_xyz_components(vcgs: Union[pandas.core.frame.DataFrame,
                                              List[pandas.core.frame.DataFrame]], legend:
                                  Optional[List[str]] = None, colours:
                                  Optional[List[List[float]]] = None, linestyle:
                                  Optional[List[str]] = None, legend_location:
                                  Optional[str] = None, limits:
                                  Optional[List[List[float]]] = None,
                                  limits_legend: Optional[List[str]] = None,
                                  limits_colours: Optional[List[List[float]]] =
                                  None, limits_linestyle: Optional[List[str]] =
                                  None, limits_legend_location: str = 'lower right',
                                  layout: str = 'grid') → tuple
```

Plot x, y, z components of VCG data

Multiple options given for layout of resulting plot

### Parameters

- **vcgs** (*list of pd.DataFrame or pd.DataFrame*) – List of vcg data: [vcg\_data1, vcg\_data2, ...]
- **legend** (*list of str, optional*) – Legend names for each VCG trace, default=None
- **colours** (*list of list of float or list of str, optional*) – Colours to use for plotting, default=common\_analysis.get\_plot\_colours
- **linestyle** (*list of str, optional*) – Linestyles to use for plotting, default='-'
- **legend\_location** (*str, optional*) – Location to plot the legend. Default=None, which will translate to 'best' if no legend is required for limits, or 'upper right' if legend is needed for limits
- **limits** (*list of list of float, optional*) – QRS limits to plot on axes, default=None To be presented in form [[qrs\_start1, qrs\_starts, ...], [qrs\_end1, qrs\_end2, ...], ...]
- **limits\_legend** (*list of str, optional*) – Legend to apply to the limits plotted, default=None
- **limits\_colours** (*list of list of float or list of str, optional*) – Colours to use when plotting limits, default=common\_analysis.get\_plot\_colours
- **limits\_linestyle** (*list of str, optional*) – Linestyles to use when plotting limits, default='-'
- **limits\_legend\_location** (*str, optional*) – Location to use for the legend containing the limits data
- **layout** (*{'grid', 'figures', 'combined', 'row', 'column', 'best'}, optional*) –

**Layout of resulting plot** grid x,y,z plots are arranged in a grid (like best, but more rigid grid) figures Each x,y,z plot is on a separate figure combined x,y,z plots are combined on a single set of axes row x,y,z plots are arranged on a horizontal row in one figure column

x,y,z plots are arranged in a vertical column in one figure best x,y,z plots are arranged to try and optimise space (nb: figures not equal sizes...)

**Returns** Handle for resulting figure(s) and axes

**Return type** fig, ax

### signalplot.vcg.plot\_xyz\_vector

`signalplot.vcg.plot_xyz_vector`(*vector: Optional[List[float]] = None, x: Optional[float] = None, y: Optional[float] = None, z: Optional[float] = None, fig: Optional[matplotlib.pyplot.figure] = None, linecolour: str = 'C0', linestyle: str = '-', linewidth: float = 2*)

Plots a specific vector in 3D space (e.g. to reflect maximum dipole)

#### Parameters

- **vector** (*list of float*) – [x, y, z] values of vector to plot, alternatively given as separate x, y, z variables
- **x** (*float*) – [x, y, z] values of vector to plot, alternatively given as vector variable
- **y** (*float*) – [x, y, z] values of vector to plot, alternatively given as vector variable
- **z** (*float*) – [x, y, z] values of vector to plot, alternatively given as vector variable
- **fig** (*plt.figure, optional*) – Existing figure handle, if desired to plot the vector onto an extant plot
- **linecolour** (*str, optional*) – Colour to plot the vector as
- **linestyle** (*str, optional*) – Linestyle to use to plot the body of the arrow
- **linewidth** (*float, optional*) – Width to plot the body of the arrow

**Returns** fig – Figure handle

**Return type** plt.figure

**Raises** **ValueError** – Exactly one of vertices and x,y,z must be given

#### Notes

Must provide either vector or [x,y,z]

## 2.3 tools

*tools.maths*

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*tools.plotting*

---

*tools.python*

---

### 2.3.1 tools.maths

#### Functions

<i>acos2</i>	Function to return the inverse cos function across the range $(-\pi, \pi]$ , rather than $(0, \pi]$
<i>asin2</i>	Function to return the inverse sin function across the range $(-\pi, \pi]$ , rather than $(-\pi/2, \pi/2]$
<i>filter_butterworth</i>	Filter data using Butterworth filter
<i>filter_savitzkygolay</i>	Filter EGM data using a Savitzky-Golay filter
<i>get_median</i>	Add the median value of data to a dataframe
<i>normalise_signal</i>	Returns a normalised signal, such that the maximum value in the signal is 1, or the minimum is -1
<i>simplex_volume</i>	Return the volume of the simplex with given vertices or sides.

#### **tools.maths.acos2**

**tools.maths.acos2**(*x*: float, *y*: float) → float

Function to return the inverse cos function across the range  $(-\pi, \pi]$ , rather than  $(0, \pi]$

##### **Parameters**

- **x** (float) – x coordinate of the point in 2D space
- **y** (float) – y coordinate of the point in 2D space

**Returns theta** – Angle corresponding to point in 2D space in radial coordinates, within range  $(-\pi, \pi]$

**Return type** float

### tools.maths.asin2

`tools.maths.asin2(x: float, y: float) → float`

Function to return the inverse sin function across the range  $(-\pi, \pi]$ , rather than  $(-\pi/2, \pi/2]$

#### Parameters

- **x** (*float*) – x coordinate of the point in 2D space
- **y** (*float*) – y coordinate of the point in 2D space

**Returns** **theta** – Angle corresponding to point in 2D space in radial coordinates, within range  $(-\pi, \pi]$

**Return type** float

### tools.maths.filter\_butterworth

`tools.maths.filter_butterworth(data: Union[numpy.ndarray, pandas.core.frame.DataFrame], sample_freq: float = 500.0, freq_filter: float = 40, order: int = 2, filter_type: str = 'low') → Union[numpy.ndarray, pandas.core.frame.DataFrame]`

Filter data using Butterworth filter

Filter a given set of data using a Butterworth filter, designed to have a specific passband for desired frequencies. It is set up to use seconds, not milliseconds.

#### Parameters

- **data** (*np.ndarray or pd.DataFrame*) – Data to filter
- **sample\_freq** (*int or float*) – Sampling rate of data (Hz), default=500. If data passed as dataframe, the sample\_freq will be calculated from the dataframe index.
- **freq\_filter** (*int or float*) – Cut-off frequency for filter, default=40
- **order** (*int*) – Order of the Butterworth filter, default=2
- **filter\_type** (*{'low', 'high', 'band'}*) – Type of filter to use, default='low'

**Returns** **filter\_out** – Output filtered data

**Return type** np.ndarray

### tools.maths.filter\_savitzkygolay

`tools.maths.filter_savitzkygolay(data: pandas.core.frame.DataFrame, window_length: int = 0.05, order: int = 2, deriv: int = 0, delta: float = 1.0)`

Filter EGM data using a Savitzky-Golay filter

Filter a given set of data using a Savitzky-Golay filter, designed to smooth data using a convolution process fitting to a low-degree polynomial within a given window. Default values are either taken from scipy documentation (not all options are provided here), or adapted to match Hermans et al.



### Parameters

- **data** (*pd.DataFrame*) – Data to filter
- **window\_length** (*float, optional*) – The length of the filter window in seconds. When passed to the scipy filter, will be converted to a positive odd integer (i.e. the number of coefficients). Default=0.05
- **order** (*int, optional*) – The order of the polynomial used to fit the samples. polyorder must be less than window\_length. Default=2
- **deriv** (*int, optional*) – The order of the derivative to compute. This must be a nonnegative integer. The default is 0, which means to filter the data without differentiating.
- **delta** (*float, optional*) – The spacing of the samples to which the filter will be applied. This is only used if deriv > 0. Default=1.0

**Returns** **data** – Output filtered data

**Return type** *pd.DataFrame*

### References

#### The development and validation of an easy to use automatic QT-interval algorithm

Hermans BJM, Vink AS, Bennis FC, Filippini LH, Meijborg VMF, Wilde AAM, Pison L, Postema PG, Delhaas T PLoS ONE, 12(9), 1–14 (2017)  
<https://doi.org/10.1371/journal.pone.0184352>

### **tools.maths.get\_median**

**tools.maths.get\_median**(*data: pandas.core.frame.DataFrame, remove\_outliers: bool = True*) → *pandas.core.frame.DataFrame*

Add the median value of data to a dataframe

TODO: Complete this code if required (currently only potentially useful for T-wave analysis)

### **tools.maths.normalise\_signal**

**tools.maths.normalise\_signal**(*data: Union[numpy.ndarray, pandas.core.frame.DataFrame]*) → *Union[numpy.ndarray, pandas.core.frame.DataFrame]*

Returns a normalised signal, such that the maximum value in the signal is 1, or the minimum is -1

**Parameters** **data** (*np.ndarray*) – Signal to be normalised

**Returns** **normalised\_data** – Normalised signal

**Return type** *np.ndarray* or *pd.DataFrame*

## tools.maths.simplex\_volume

`tools.maths.simplex_volume(*, vertices=None, sides=None) → float`

Return the volume of the simplex with given vertices or sides.

If vertices are given they must be in a NumPy array with shape (N+1, N): the position vectors of the N+1 vertices in N dimensions. If the sides are given, they must be the compressed pairwise distance matrix as returned from `scipy.spatial.distance.pdist`.

Raises a `ValueError` if the vertices do not form a simplex (for example, because they are coplanar, colinear or coincident).

Warning: this algorithm has not been tested for numerical stability.

## 2.3.2 tools.plotting

### Functions

<code>add_colourbar</code>	Add arbitrary colourbar to a figure, for instances when an automatic colorbar isn't available
<code>add_xyz_axes</code>	Plot dummy axes (can't move splines in 3D plots)
<code>get_plot_colours</code>	Return iterable list of RGB colour values that can be used for custom plotting functions
<code>get_plot_lines</code>	Returns different line-styles for plotting
<code>set_axis_limits</code>	Set axis limits (not automatic for line collections, so needs to be done manually)
<code>set_symmetrical_axis_limits</code>	Sets symmetrical limits for a series of axes
<code>write_colourmap_to_xml</code>	Create a Paraview friendly colourmap useful for highlighting a particular range

## tools.plotting.add\_colourbar

`tools.plotting.add_colourbar(limits: List[float], fig: Optional[matplotlib.pyplot.figure] = None, colourmap: str = 'viridis', n_elements: int = 100) → None`

Add arbitrary colourbar to a figure, for instances when an automatic colorbar isn't available

### Parameters

- **limits** (*list of float*) – Numerical limits to apply
- **fig** (*plt.figure, optional*) – Figure on which to plot the colourbar. If not provided (default=None), then will pick up the figure most recently available
- **colourmap** (*str, optional*) – Colourmap to be used, default='viridis'
- **n\_elements** (*int, optional*) – Number of entries to be made in the colourmap index, default=100

## Notes

This is useful for instances such as when LineCollections are used to plot line that changes colour during the plotting process, as LineCollections do not enable an automatic colorbar to be added to the plot. This function adds a dummy colorbar to replace that.

## tools.plotting.add\_xyz\_axes

`tools.plotting.add_xyz_axes`(*fig: matplotlib.pyplot.figure, ax: mpl\_toolkits.mplot3d.axes3d.Axes3D, axis\_limits: Optional[Union[float, List[float], List[List[float]]] = None, symmetrical\_axes: bool = False, equal\_limits: bool = False, unit\_axes: bool = False, sig\_fig: Optional[int] = None*) → None

Plot dummy axes (can't move splines in 3D plots)

### Parameters

- **fig** (*plt.figure*) – Figure handle
- **ax** (*Axes3D*) – Axis handle
- **axis\_limits** (*float or list of float or list of list of float, optional*) – Axis limits, either same for all dimensions (min=-max), or individual limits ([min, max]), or individual limits for each dimension
- **symmetrical\_axes** (*bool, optional*) – Apply same limits to x, y and z axes
- **equal\_limits** (*bool, optional*) – Set axis minimum to minus axis maximum (or vice versa)
- **unit\_axes** (*bool, optional*) – Apply minimum of -1 -> 1 for axis limits
- **sig\_fig** (*int, optional*) – Maximum number of decimal places to be used on the axis plots (e.g., if set to 2, 0.12345 will be displayed as 0.12). Used to avoid floating point errors, default=None (no adaption made)

## tools.plotting.get\_plot\_colours

`tools.plotting.get_plot_colours`(*n: int = 10, colourmap: Optional[str] = None*) → List[Tuple[float]]

Return iterable list of RGB colour values that can be used for custom plotting functions

Returns a list of RGB colours values, potentially according to a specified colourmap. If *n* is low enough, will use the custom 'tab10' colourmap by default, which will use alternating colours as much as possible to maximise visibility. If *n* is too big, then the default setting is 'viridis', which should provide a gradation of colour from first to last.

### Parameters

- **n** (*int, optional*) – Number of distinct colours required, default=10
- **colourmap** (*str*) – Matplotlib colourmap to base the end result on. Will default to 'tab10' if *n*<11, 'viridis' otherwise

**Returns** `cmap` – List of RGB values

**Return type** list of tuple

### `tools.plotting.get_plot_lines`

`tools.plotting.get_plot_lines(n: int = 4) → Union[List[tuple], List[str]]`

Returns different line-styles for plotting

**Parameters** `n` (*int*, *optional*) – Number of different line-styles required

**Returns** `lines` – List of different line-styles

**Return type** list of str or list of tuple

### `tools.plotting.set_axis_limits`

`tools.plotting.set_axis_limits(ax, data: Optional[pandas.core.frame.DataFrame] = None, unit_min: bool = True, axis_limits: Optional[Union[float, List[float]]] = None, pad_percent: float = 0.01) → None`

Set axis limits (not automatic for line collections, so needs to be done manually)

#### **Parameters**

- **ax** – Handles to the axes that need to be adjusted
- **data** (*pd.DataFrame*, *optional*) – Data that has been plotted, default=None
- **unit\_min** (*bool*, *optional*) – Whether to have the axes set to, as a minimum, unit length, default=True
- **axis\_limits** (*list of float or float*, *optional*) – Min/max values for axes, either as one value (i.e. min=-max), or two separate values. Same axis limits will be applied to all dimensions
- **pad\_percent** (*float*, *optional*) – Percentage ‘padding’ to add to the ranges, to try and ensure that the edges of linewidths are not cut off, default=0.01

### `tools.plotting.set_symmetrical_axis_limits`

`tools.plotting.set_symmetrical_axis_limits(ax_min: float, ax_max: float, unit_axes: bool = False) → Tuple[float, float]`

Sets symmetrical limits for a series of axes

TODO: fold functionality into `set_axis_limits` to avoid redundant functions

#### **Parameters**

- **ax\_min** (*float*) – Minimum value for axes
- **ax\_max** (*float*) – Maximum value for axes
- **unit\_axes** (*bool*, *optional*) – Whether to apply a minimum axis range of [-1,1]

Returns **ax\_min, ax\_max** – Symmetrical axis limits, where  $ax\_min = -ax\_max$

Return type float

### tools.plotting.write\_colourmap\_to\_xml

`tools.plotting.write_colourmap_to_xml(start_data: float, end_data: float, start_highlight: float, end_highlight: float, opacity_data: float = 1, opacity_highlight: float = 1, n_tags: int = 20, colourmap: str = 'viridis', outfile: str = 'colourmap.xml') → None`

Create a Paraview friendly colourmap useful for highlighting a particular range

Creates a colourmap that is entirely gray, save for a specified region of interest that will vary according to the specified colourmap

**start\_data** start value for overall data (can't just use data for region of interest - Paraview will scale)

**end\_data** end value for overall data **start\_highlight** start value for region of interest **end\_highlight** end value for region of interest

**opacity\_data** 1.0 overall opacity to use for all data **opacity\_highlight** 1.0 opacity for region of interest **colourmap** 'viridis' colourmap to use **outfile** 'colourmap.xml' filename to save .xml file under

None

## 2.3.3 tools.python

### Functions

<code>check_list_depth</code>	Function to calculate the depth of nested loops
<code>convert_index_to_time</code>	Return 'real' time for a given index
<code>convert_input_to_list</code>	Convert a given input to a list of inputs of required length.
<code>convert_time_to_index</code>	Converts a given time point to the relevant index value
<code>deprecated_convert_time_to_index</code>	Return indices of QRS start and end points.
<code>find_list_fraction</code>	Find index corresponding to certain fractional length within a list, e.g.
<code>get_i_colour</code>	Get index appropriate to colour value to plot on a figure (will be 0 if brand new figure)
<code>get_time</code>	Returns variables for time, dt and t_end, depending on input.
<code>recursive_len</code>	Return the total number of elements with a potentially nested list

### tools.python.check\_list\_depth

`tools.python.check_list_depth(input_list, depth_count=1, max_depth=0, n_args=0)`  
 Function to calculate the depth of nested loops

TODO: Finish this damn code

#### Parameters

- **input\_list** (*list*) – Input argument to check
- **depth\_count** (*int, optional*) – Depth of nested loops thus far
- **max\_depth** (*int, optional*) – Maximum expected depth of list, default=0 (not checked)
- **n\_args** (*int, optional*) – Required length of ‘base’ list, default=0 (not checked)

**Returns** **depth\_count** – Depth of nested loops

**Return type** **int**

#### Notes

A list of form [a1, a2, a3, ...] has depth 1. A list of form [[a1, a2, a3, ...], [b1, b2, b3, ...], ...] has depth 2. And so forth...

If n\_args is set to an integer greater than 0, it will check that the lowest level of lists (for all entries) will be of the required length

if depth=1 as above, `len([a1, a2, a3, ...]) == n_args` if depth=2 as above, `len([a1, a2, a3, ...]) == n_args && len([b1, b2, b3, ...]) == n_args`

### tools.python.convert\_index\_to\_time

`tools.python.convert_index_to_time(idx: int, time: Optional[numpy.ndarray] = None, t_start: float = 0, t_end: float = 200, dt: float = 2) → float`

Return ‘real’ time for a given index

#### Parameters

- **idx** (*int*) – Index to convert
- **time** (*np.ndarray, optional*) – Time data; if not provided, will be assumed from t\_start, t\_end and dt variables, default=None
- **t\_start** (*float, optional*) – Start time for overall data, default=0
- **t\_end** (*float, optional*) – End time for overall data, default=200
- **dt** (*float, optional*) – Interval between time points, default=2

**Returns** **time** – The time value that corresponds to the given index

**Return type** **float**

## tools.python.convert\_input\_to\_list

`tools.python.convert_input_to_list(input_data: Any, n_list: int = 1, n_list2: int = -1, list_depth: int = 1, default_entry: Optional[Any] = None) → list`

Convert a given input to a list of inputs of required length. If already a list, will confirm that it's the right length.

### Parameters

- **input\_data** (*Any*) – Input argument to be checked
- **n\_list** (*int, optional*) – Number of entries required in input; if set to -1, will not perform any checks beyond 'depth' of lists, default=1
- **n\_list2** (*int, optional*) – Number of entries for secondary input; if set to -1, will not perform any checks
- **list\_depth** (*int*) – Number of nested lists required. If just a simple list of e.g. VCGs, then will be 1 ([vcg1, vcg2,...]). If a list of lists (e.g. [[qrs\_start1, qrs\_start2,...], [qrs\_end1, qrs\_end2,...]), then 2.
- **default\_entry** (*{'colour', 'line', None, Any}, optional*) – Default entry to put into list. If set to None, will just repeat the input data to match n\_list. However, if set to either 'colour' or 'line', will return the default potential settings, default=None

**Returns** **output** – Formatted output

**Return type** list

### Notes

If the data are already provided as a list and list\_depth==1, function will simply check that the list is of the correct length. If list\_depth==2, will check that deepest level of nesting has the correct length; if n\_list2 is provided, it will check the top level of the list is of the correct length. This is used, for example, when several different limits are provided for several different VCGs, and a legend is needed. Thus, if there are n different VCGs to be plotted, and each has m different limits to be plotted, the legend can be checked to be of the form [[x11, x21,...,xn1], [x12, x22,...,xn2],...[x1m, x2m,...,xnm]]

**If the data are not in list form, will:**

- if default\_entry==None, will replicate input\_data to match n\_vcg, e.g. '-' becomes ['-','-',...]
- if default\_entry=='colour', will return list of RGB values for colours
- if default\_entry=='line', will return list of line entries
- for any other value of default\_entry, will reproduce that value

### tools.python.convert\_time\_to\_index

```
tools.python.convert_time_to_index(time_point: float, time:
                                   Optional[Union[List[float], numpy.ndarray]] =
                                   None, t_start: Optional[float] = None, t_end:
                                   Optional[float] = None, dt: Optional[float] =
                                   None) → int
```

Converts a given time point to the relevant index value

#### Parameters

- **time\_point** (*float*) – Time point for which we wish to find the corresponding index. If set to -1, will return the final index
- **time** (*float or np.ndarray, optional*) – Time data from which we wish to extract the index. If set to None, the time will be constructed based on the assumed t\_start, t\_end and dt values
- **t\_start** (*float, optional*) – Start point of time; only used if 'time' variable not given, default=None
- **t\_end** (*float, optional*) – End point of time; only used if 'time' variable not given, default=None
- **dt** (*float, optional*) – Interval between time points; only used if time not given, default=None

**Returns** **i\_time** – Index corresponding to the time point given

**Return type** int

**Raises** **AssertionError** – If insufficient data are provided to the function to enable it to function

### tools.python.deprecated\_convert\_time\_to\_index

```
tools.python.deprecated_convert_time_to_index(qrs_start: Optional[float] = None,
                                              qrs_end: Optional[float] = None,
                                              time: Optional[List[float]] =
                                              None, t_start: float = 0, t_end:
                                              float = 200, dt: float = 2) →
                                              Tuple[int, int]
```

Return indices of QRS start and end points. NB: indices returned match Matlab output

**..deprecated::** This function is deprecated, but is in use due to other functions still using it for the moment

#### Parameters

- **qrs\_start** (*float or int, optional*) – Start time to convert to index. If not given, will default to the same as the start time of the entire list
- **qrs\_end** (*float or int, optional*) – End time to convert to index. If not given, will default to the same as the end time of the entire list
- **time** (*float, optional*) – Time data to be used to calculate index. If given, will over-ride the values used for dt/t\_start/t\_end. Default=None



- **t\_start** (*float or int, optional*) – Start time of overall data, default=0
- **t\_end** (*float or int, optional*) – End time of overall data, default=200
- **dt** (*float or int, optional*) – Interval between time points, default=2

#### Returns

- **i\_qrs\_start** (*int*) – Index of start time
- **i\_qrs\_end** (*int*) – Index of end time

### tools.python.find\_list\_fraction

`tools.python.find_list_fraction(input_list, fraction=0.5, interpolate=True)`

Find index corresponding to certain fractional length within a list, e.g. halfway along, a third along

If only looking for an interval halfway along the list, uses a simpler method that is computationally faster

`input_list` List to find the fractional value of

`fraction` 0.5 Fraction of length of list to return the value of `interpolate` True If fraction does not precisely specify a particular entry in the list, whether to return the

values on either side, or whether to interpolate between the two values (with weighting given to how close the fraction is to one value or the other)

### tools.python.get\_i\_colour

`tools.python.get_i_colour(axis_handle) → int`

Get index appropriate to colour value to plot on a figure (will be 0 if brand new figure)

### tools.python.get\_time

`tools.python.get_time(time: Optional[numpy.ndarray] = None, dt: Optional[float] = None, t_end: Optional[float] = None, n_vcg: Optional[int] = 1, len_vcg: Optional[List[int]] = None) → Tuple[List[numpy.ndarray], List[float], List[float]]`

Returns variables for time, dt and t\_end, depending on input.

#### Parameters

- **time** (*np.ndarray, optional*) – Time data for a given VCG, default=None
- **dt** (*float, optional*) – Interval between recording points for the VCG, default=None
- **t\_end** (*float, optional*) – Total duration of the VCG recordings, default=None
- **n\_vcg** (*int, optional*) – Number of VCGs being assessed, default=1

- **len\_vcg** (*int, optional*) – Number of data points for each VCG being assessed, None

#### Returns

- **time** (*list of np.ndarray*) – Time data for a given VCG
- **dt** (*list of float*) – Mean time interval for a given VCG recording
- **t\_end** (*list of float*) – Total duration of each VCG recording

#### Notes

Time OR t\_end/dt/len\_vcg must be passed to this function

#### **tools.python.recursive\_len**

`tools.python.recursive_len(item: list)`

Return the total number of elements with a potentially nested list

New stuff7

## INDICES AND TABLES

Note that, at the moment, none of the following links actually work...

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