# Timeseries II

10.11.24

## **Learning Objectives**

Date-Time Functions

Moving Windows

Fourier

#### **Pandas Dates**

- Extensive handling of date-times
- Timestamps: singular moments
  - Can have any spacing between timestamps
- Periods: a duration of time, e.g. first-quarter
  - Periods are regularly spaced (same time in each period)
- Choice doesn't really affect most Pandas functions.

#### Pandas date initialization with date\_range

Create 10 days ("D") starting in October 2024:

```
Oct=pd.date_range("2024-10",\
periods=10,freq="D")
```

```
DatetimeIndex(['2024-10-01', '2024-10-02', '2024-10-03', '2024-10-04',
'2024-10-05', '2024-10-06', '2024-10-07', '2024-10-08',
'2024-10-09', '2024-10-10'],
dtype='datetime64[ns]', freq='D')
```

#### Pandas date initialization with date\_range

Pass to create some periodic data:

```
Oct=pd.date_range("2024-10",\
periods=10,freq="D")

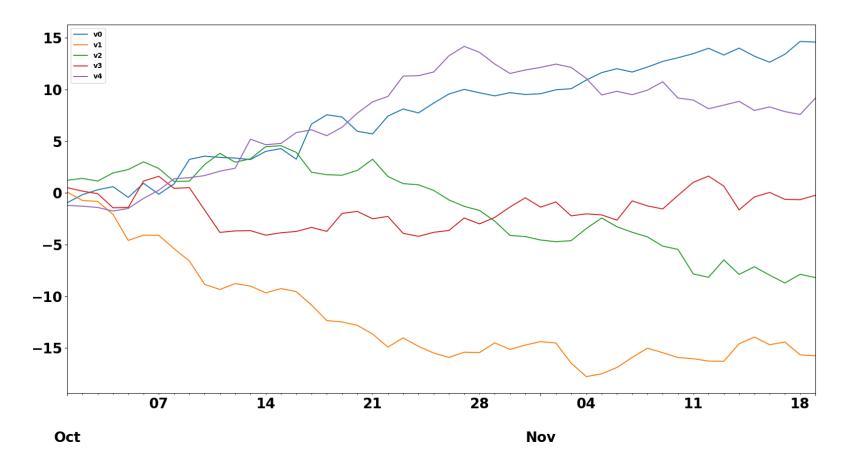
TimeData=pd.DataFrame(data,columns=cNames,index=Oct);
```

Can now select data based upon time-ranges:

```
TimeData['2024-10-01':'2024-10-09']
```

#### Pandas timeseries plot

Directly call plot method from timeseries dataframe: TimeData.plot()



#### Pandas resample

- You can downsample timeseries using aggregate rules
- e.g. give min value every 3 days:

```
dnSamp=TimeData.resample('3D').min();
```

• Other aggregators: max, mean, median etc.

## Convolution: Reminder

#### Example 1

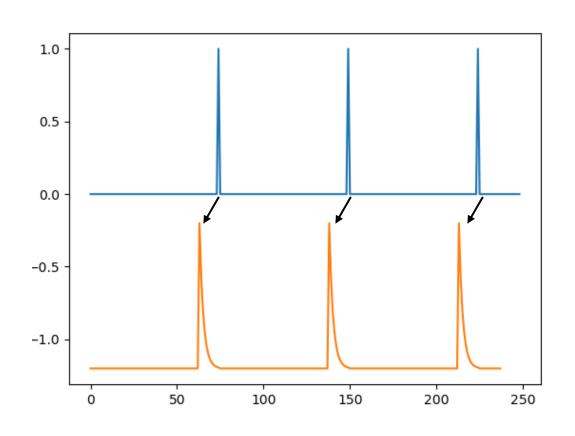
- Step 1: **flip** *g*
- Step 2: moving sum

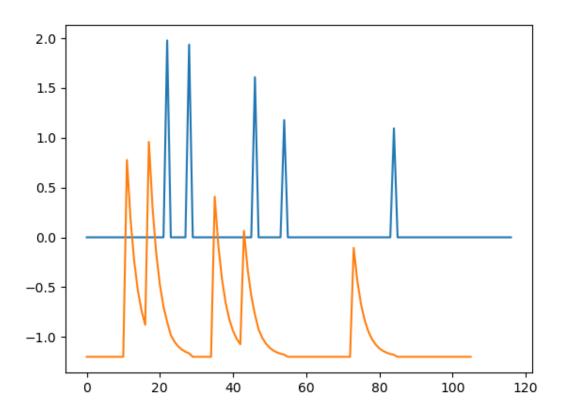
В 3A+B [A В E] 3B+C В [A  $\mathsf{E}]$ 3C+D [A 3D+E

f : [A B C D E]

g: [1 3]

### Point spread as Spike-Responses





np.convolve(data, kernel, 'valid')

#### **Manual Convolution**

np.convolve(data,kern,type)

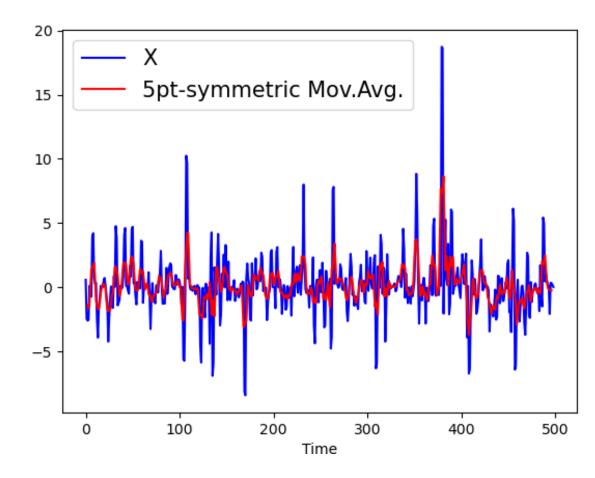
**Start End** type='valid': Zero-padding type='same': 3 type='full':

#### 5-point moving average (on-center)

np.convolve(data,kern,type)

Shifted values of X to account for kernel

$$y_t = \frac{1}{5} \sum_{k=-2}^{2} x_{t+k}$$



#### Pandas Rolling

 Use aggregate fcts over moving-windows

#### 3 column data-frame: data

```
1
-0.159214
           3.628687 -0.565338
                     4.445897
10.735521
           5.898765
                    -0.233595
10.843167
           0.319493
5.757754
           4.330009
                     3.834718
5.824845
           1.074398
                     3.548873
7.182152
           3.595056
                      6.071613
 3.169256
           5.210517
                      8.640147
 5.400332
                      4.475311
           4.420175
```

- Window length: 5
- Start at position 3 (else NaN)

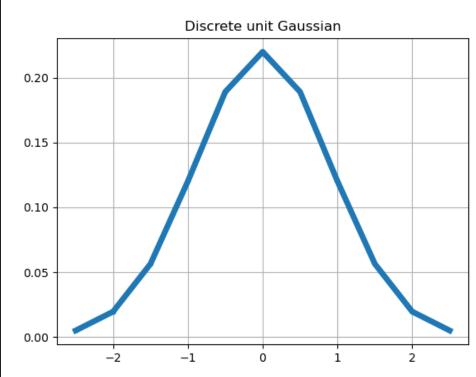
#### data.rolling(5,min\_periods=3).max()

	0	1	2
0	NaN	NaN	NaN
1	NaN	NaN	NaN
2	10.843167	5.898765	4.445897
3	10.843167	5.898765	4.445897
4	10.843167	5.898765	4.445897
5	10.843167	5.898765	6.071613
6	10.843167	5.210517	8.640147
7	7.182152	5.210517	8.640147

#### Pandas Rolling: Gaussian filter

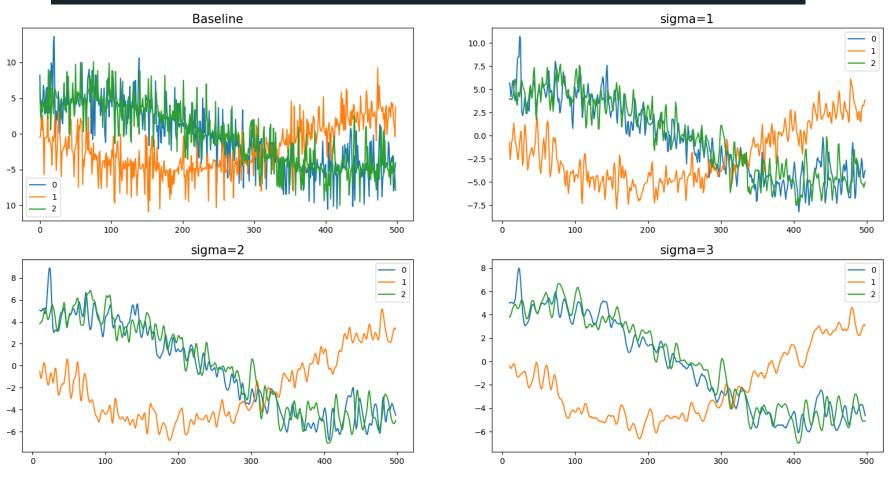
On-center Gaussian with variable SD

```
def GaussFilt(X, sigma):
    ## Setup some parameters
    var2=2*(sigma**2);
    sz=X.shape[0];
   ## Shift x-values on-center, variable length
    xVals=np.linspace(-sz/2,sz/2,sz);
    ## Discrete Gaussian
    gKern=np.exp(-(xVals**2)/var2);
    gKern=gKern/np.sum(gKern);
    ## Inner product
    return gKern.reshape([1,sz])@X,gKern,xVals
```



#### Pandas Rolling: Gaussian Filter

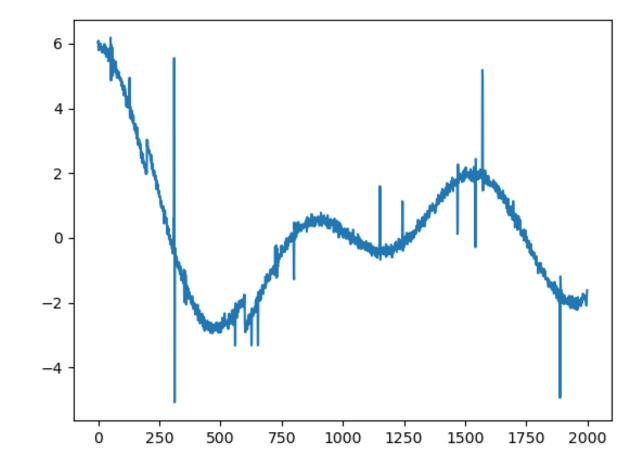
mvGauss=data.rolling(11).apply\
 (lambda x:GaussFilt(x,sigma=kk+1)[0])



#### Practice Together: Median Filter

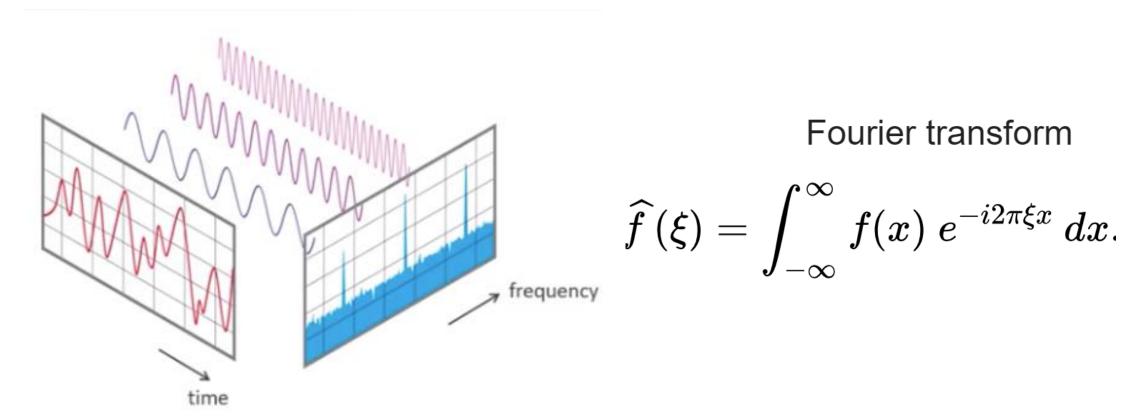
• I have a timeseries with both drift and spikes:

• I need to detrend this signal



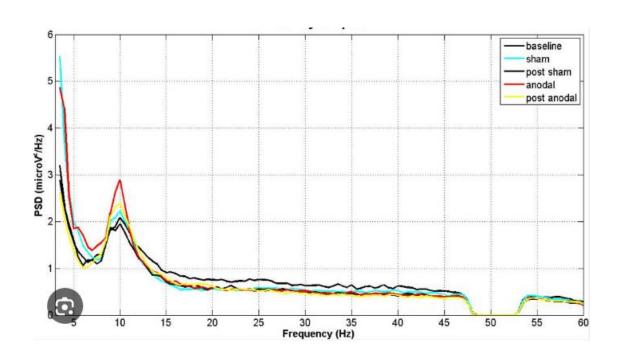
#### Fourier Analysis: From time to frequency

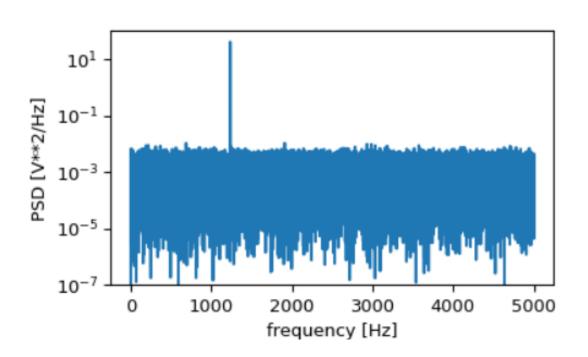
- NumPy fft submodule
- Many functions in the scipy signals submodule



## Power-Spectral Density: signal.periodogram(x,Fs)

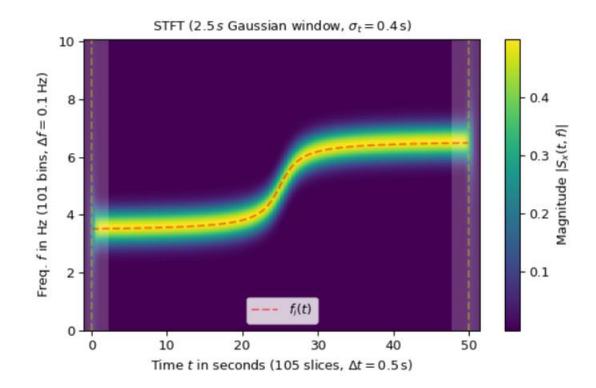
- How much is there of each signal?
- Plotted as power (squared complex-mod) of FT





#### Spectrograms

- Short-time frequency content
- SciPy spectrogram of an oscillator shifting from 3.5 to 6.5 Hz



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# Fin