

# Time

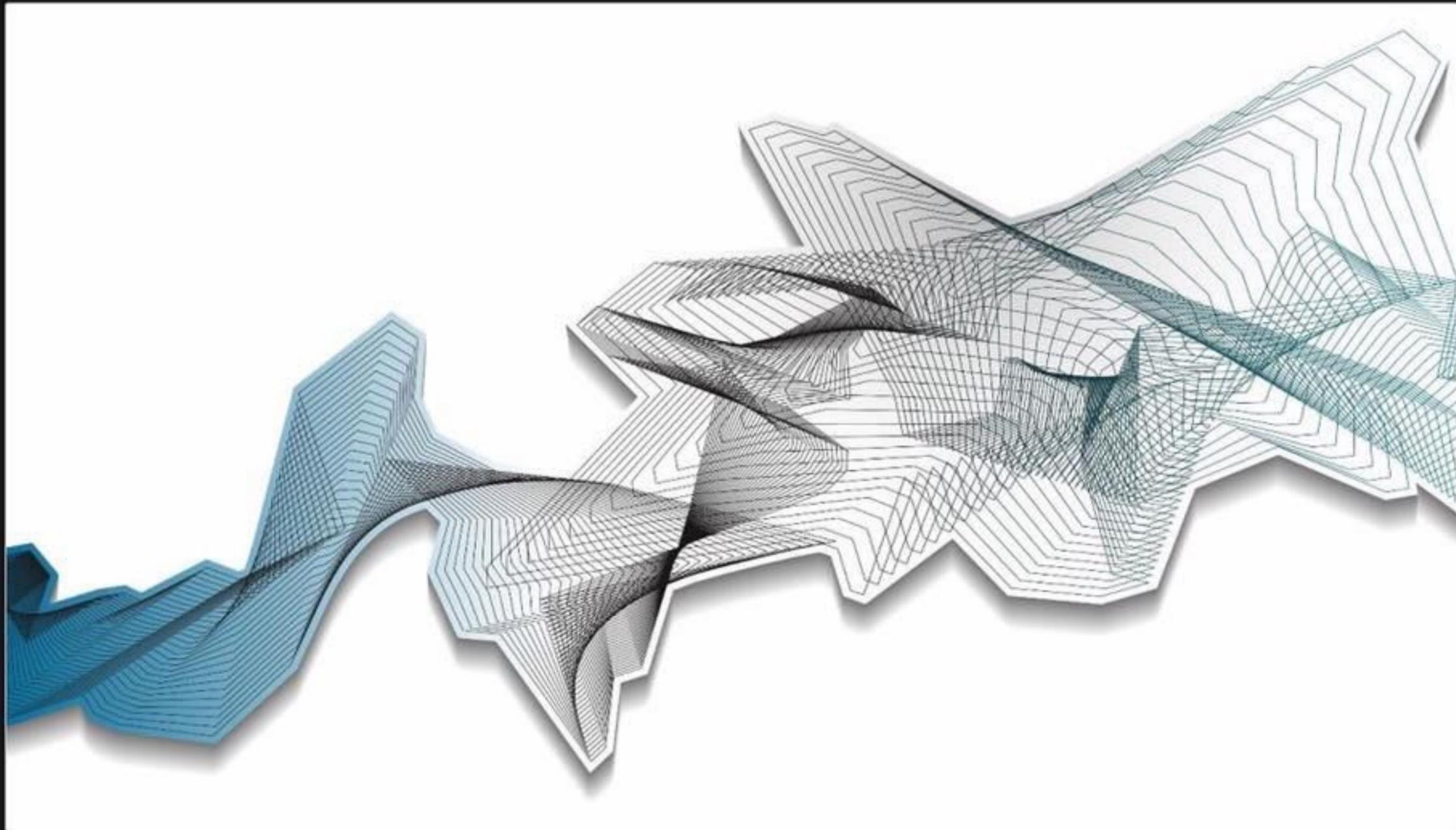
Raoul Grouls, 5 maart 2024

# Leerdoelen les 3

- Leren toepassen van visualisatie principes
- Omgaan met venv, pdm, path, scripts, git
- Werken met timestamps
- Autocorrelation (statsmodels)
- Seasonal decompose (Trend, Seasonal, Residu)
- Fourier transforms: the main idea
- Using Fourier transforms to model timeseries
- Python
  - Pandas .dt (isoweek, date, day\_name, see docs for more)
  - Pandas reindex
  - Seaborn FacetGrid
  - Seaborn .map
  - Plotly px.area
  - Statsmodels acf, seasonal\_decompose
  - scipy.fft, scipy.signal

# Git

- The only sane way to work together on code
- <https://learngitbranching.js.org/>



# Version Control

Git gets easier once you get the basic idea that branches are homeomorphic endofunctors mapping submanifolds of a Hilbert space.

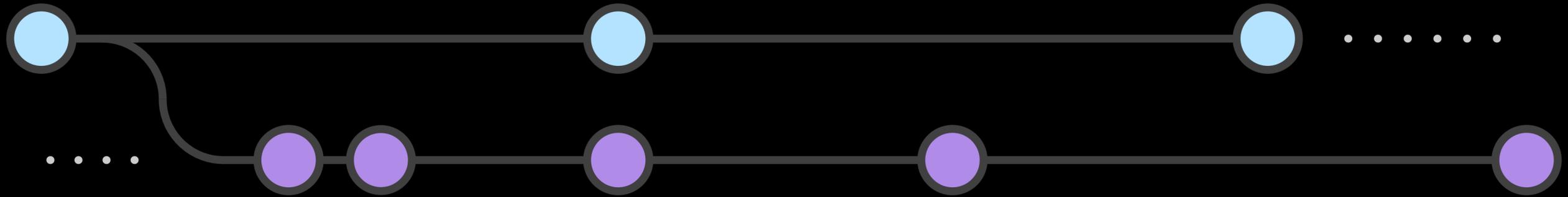
Main

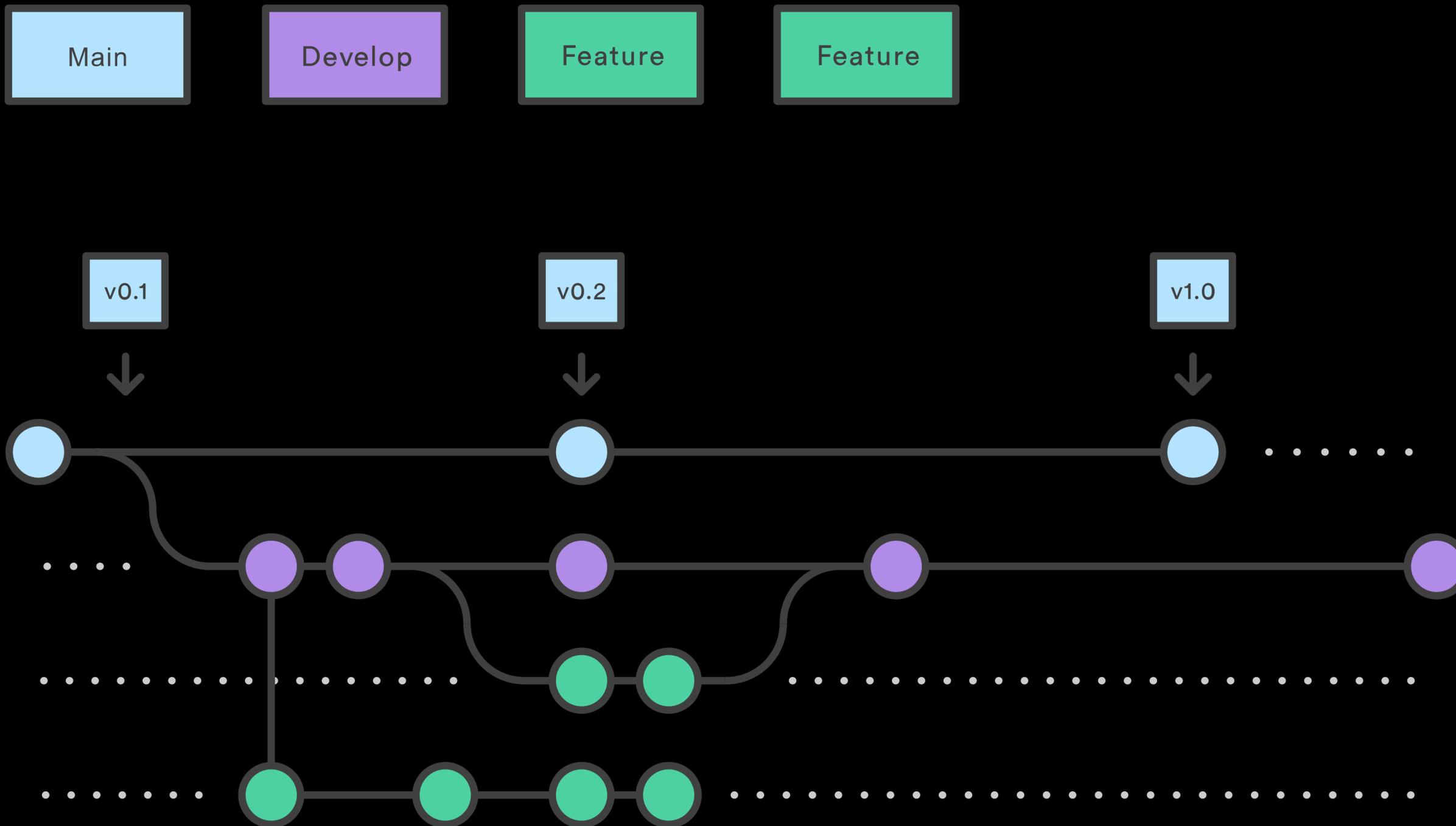
Develop

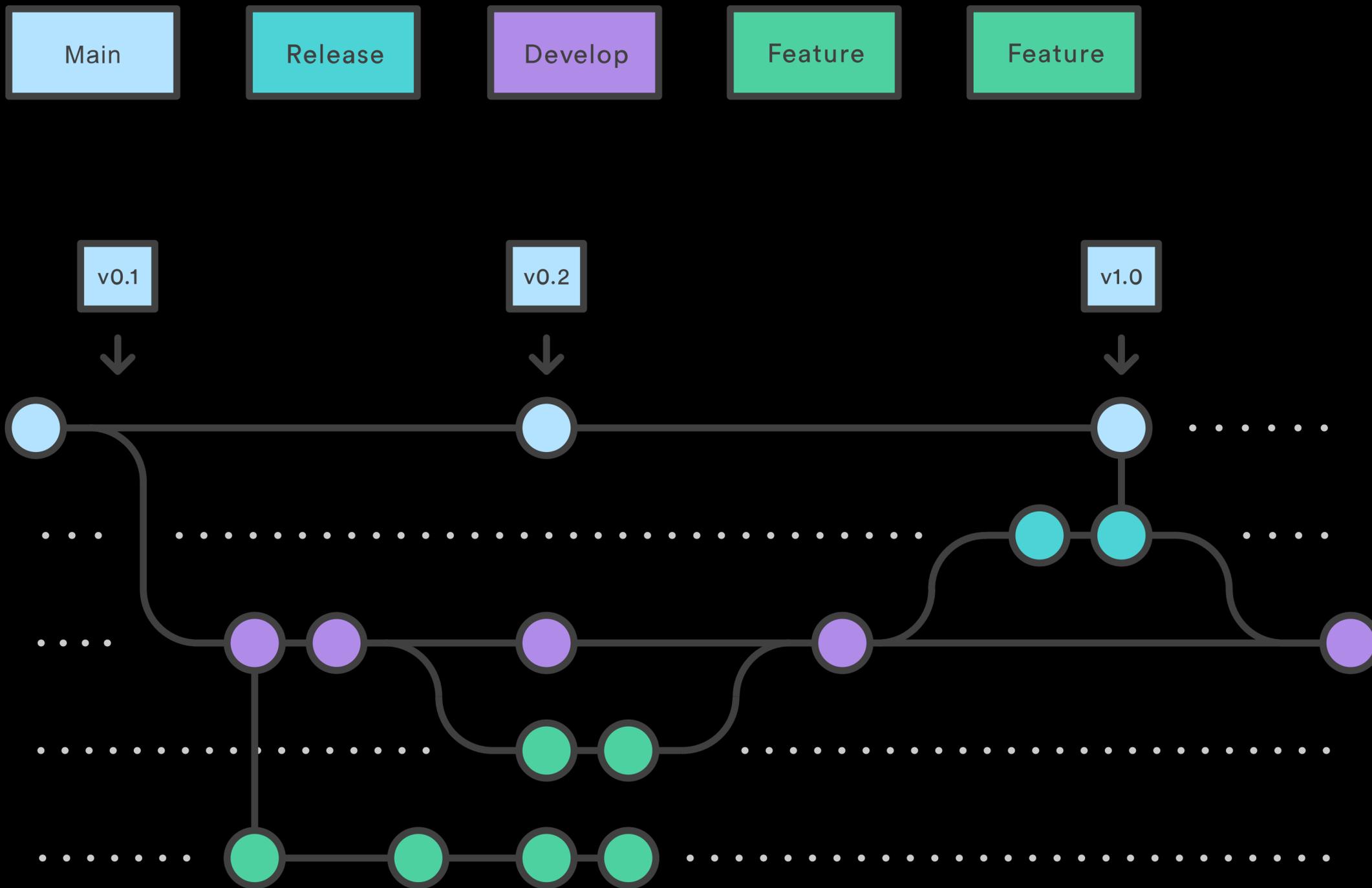
v0.1

v0.2

v1.0





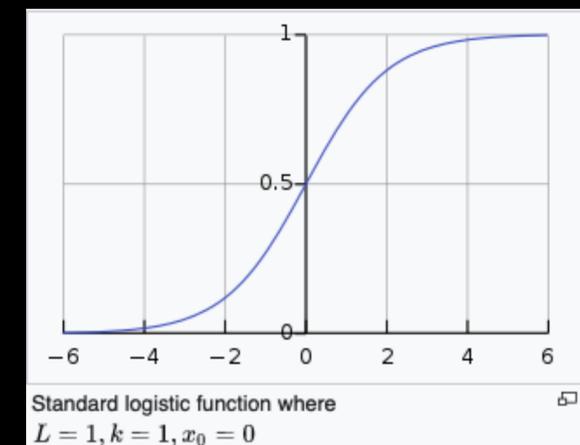
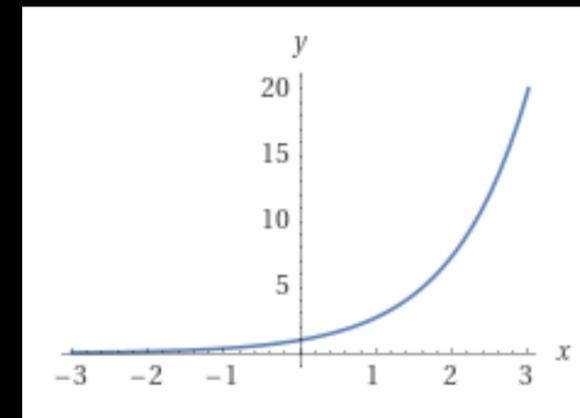
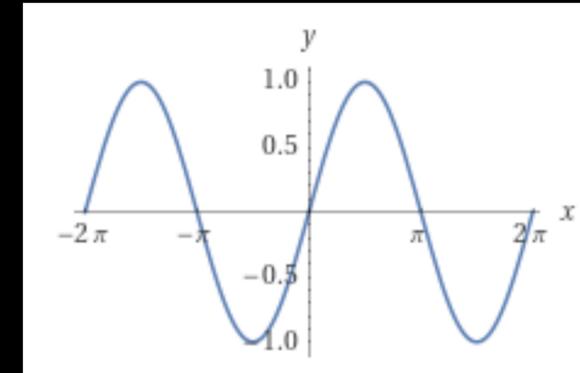


# The four horsemen of modelling

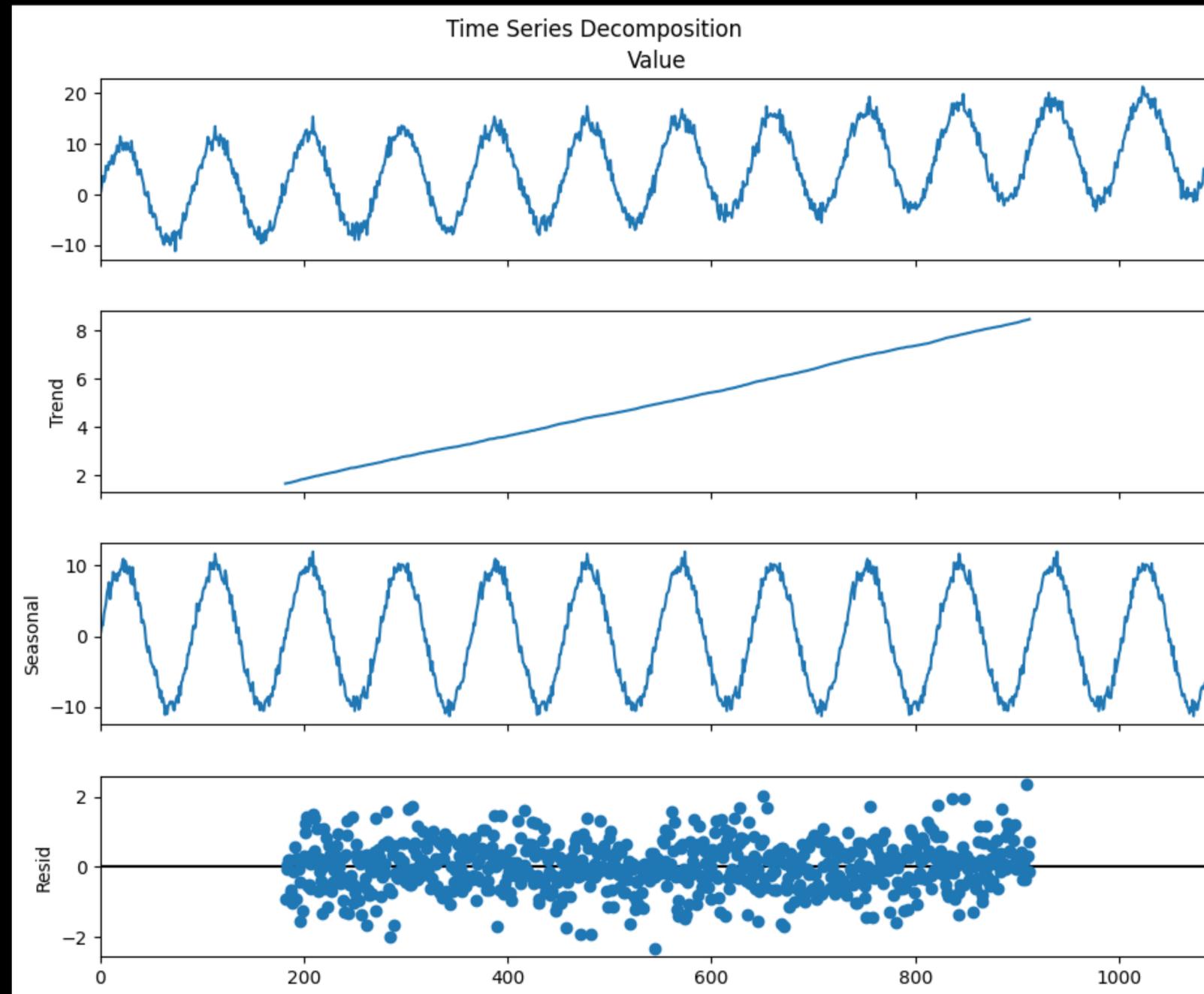
- Linear
- Sine
- Logistic
- Exponential

# The four horsemen of modelling

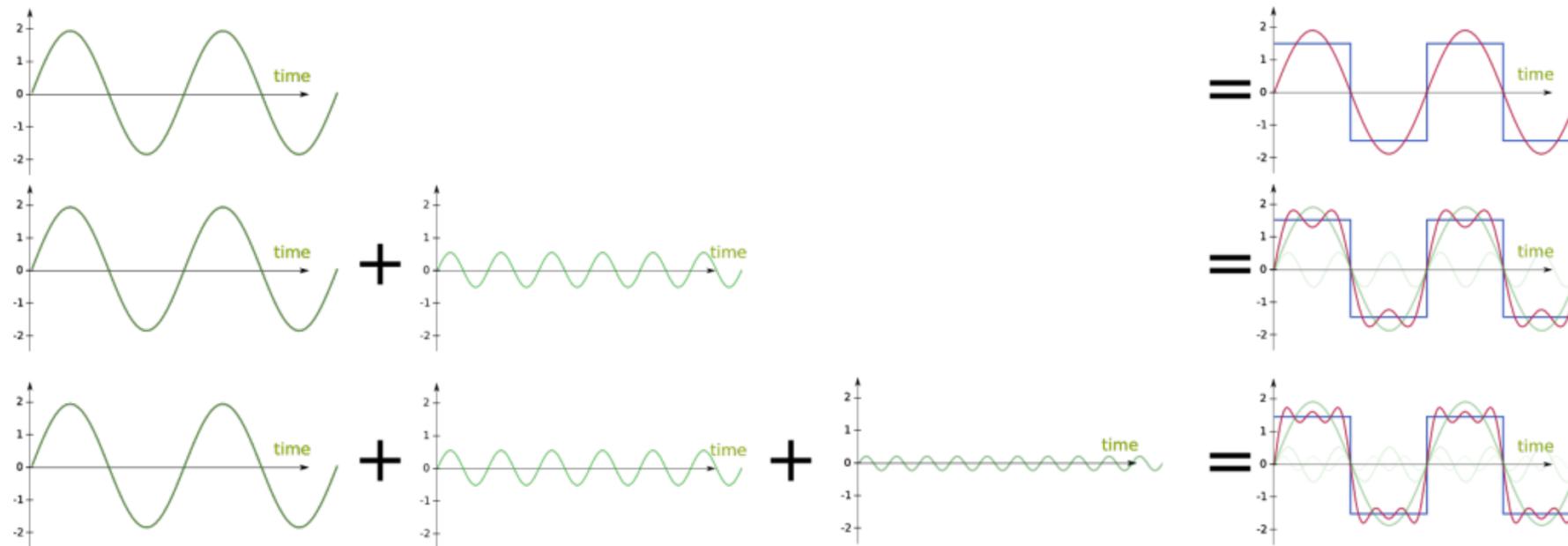
- Linear:  $f(X) = WX + b$
- Sine:  $f(t) = A \cdot \sin(\omega t + \phi)$  with A for amplitude,  $\omega$  for angular frequency (radians/sec), and  $\phi$  for phase shift with  $0 \leq \phi \leq 2\pi$
- Exponential:  $f(x) = e^x$
- Logistic:  $f(x) = \frac{L}{1 + e^{-k(x-x_0)}}$  with L max value, k growth rate and  $x_0$  midpoint



# Timeseries decomposition



# Fourier Transforms



And so on...

# Fourier Transforms

