

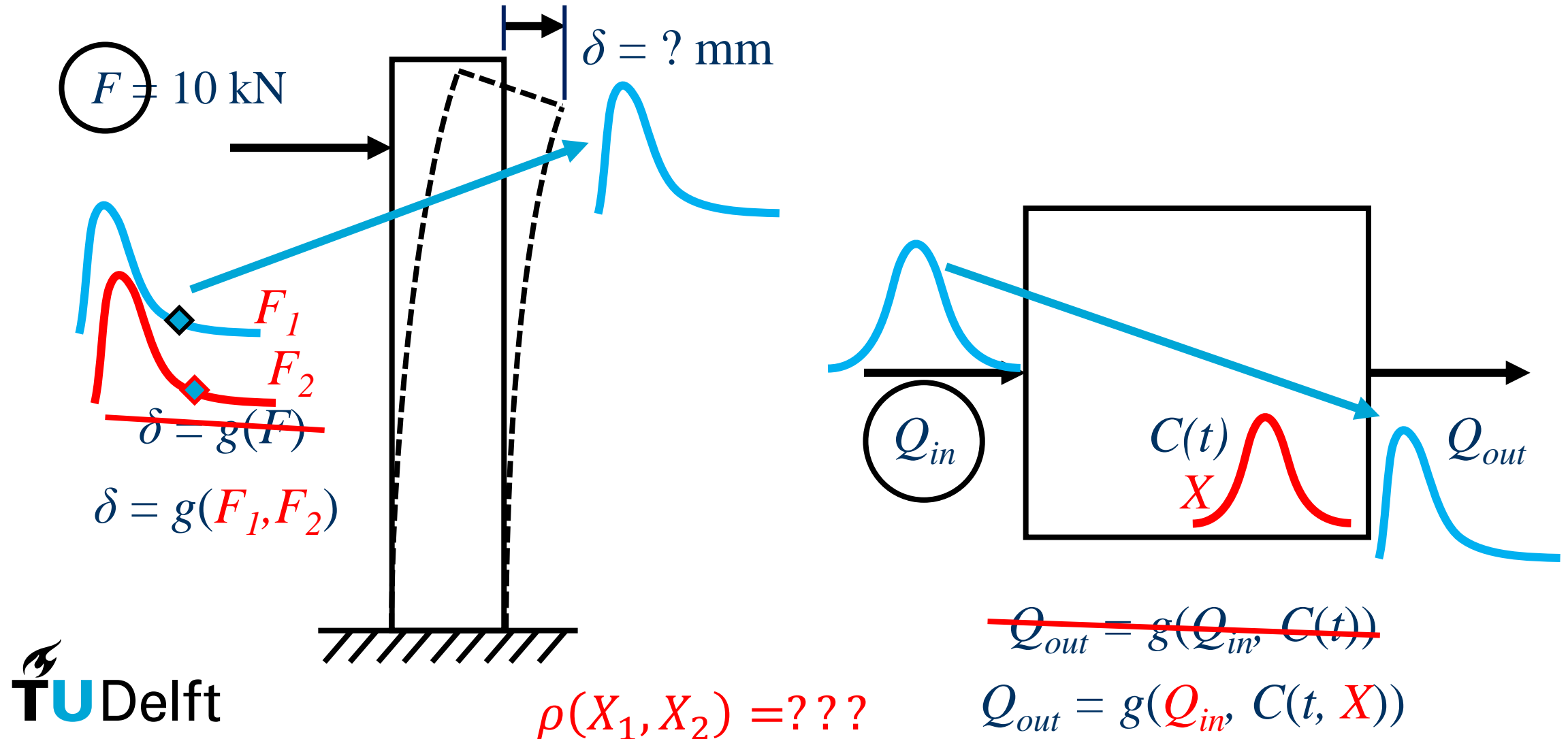
# Modelling, Uncertainty and Data for Engineers (MUDE)

## Week 1.8 : Multivariate Distributions

There was a lot of work on the board in this lecture. Some slides have been added as a supplement/quick reference (with a box like this), but the recording should be watched to get the complete story; a similar example to the island case here is worked out in the textbook (Chapter 6).

<https://collegeramavideoportal.tudelft.nl/catalogue/cegm1000>

# Incorporating (More!) Uncertainty in our Models



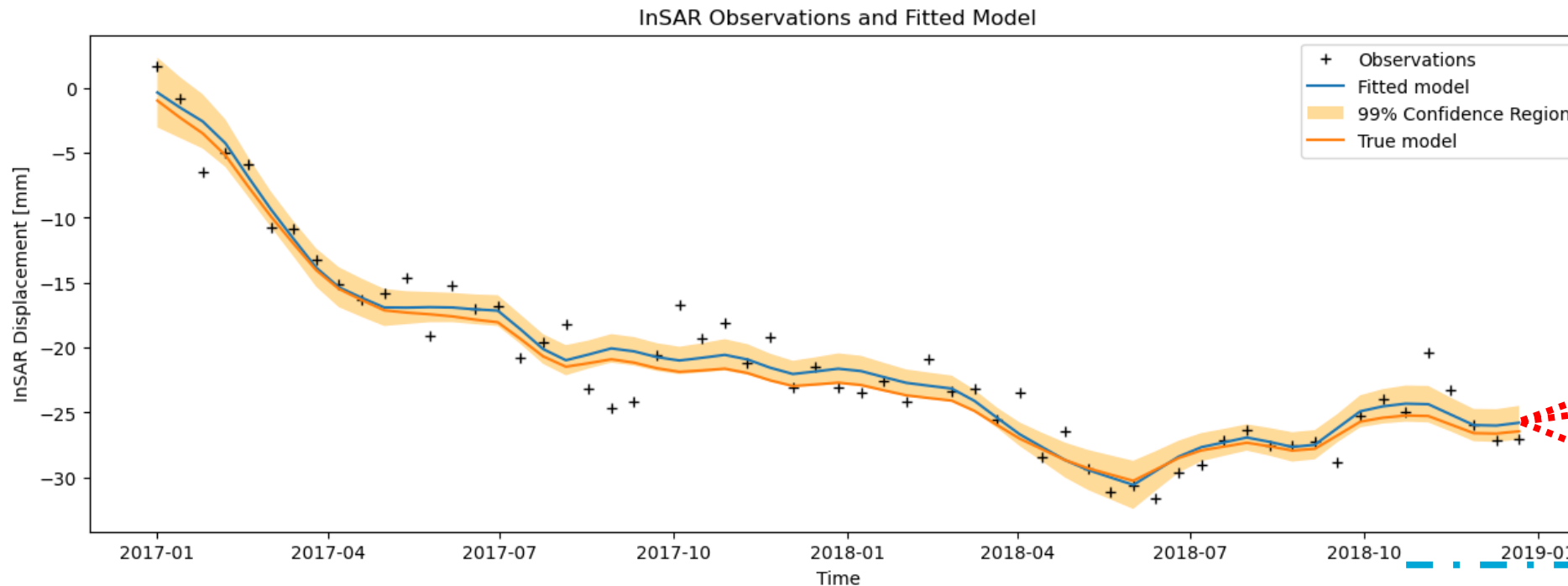
# Project 2

$$d = d_0 + vt + k \text{ GW}$$

- $d_0$ : initial displacement at  $t_0$
- $v$ : displacement velocity
- $k$ : soil response to change in GW

$$d = d_0 + R(1 - \exp - \frac{t}{a}) + k \text{ GW}$$

- $d_0$ : initial displacement at  $t_0$
- $R$ : soil response to road
- $k$ : soil response to change in GW



To use model, we need:

- Multivariate model, incorporating probability
- Non-Gaussian Dist.'s
- Non-linear functions

^Threshold?

$P(d \text{ exceeds threshold}) = ?$

# Today: tools for creating a multivariate model

- Covariance, dependence
  - Functions of random variables
  - Monte Carlo simulation
- 
- We've already done these things before!
  - A bit of context + math to get more comfortable with probability

# Why do we want to use probability?

## A simple example

# See Collegerama Recording

Oceanic island example:

- Coral atoll, susceptible to sea level rise and settlement due to dissolution of coral (calcium carbonate)
  - Current elevation = 10 m + mean sea level (MSL)
  - Settlement,  $D \sim N(\mu = 3\text{m}, \sigma = 1.0 \text{ m})$
  - Sea level rise,  $S \sim N(\mu = 5\text{m}, \sigma = 1.6 \text{ m})$
  - Observations: from a few atolls around the world (440 in total)
  - Distributions: prediction of what will happen over the next 100 year period
- 
- How can we evaluate the probability that the island is underwater?

*See Chapter 6 in textbook for similar example*

# Settlement Due to Ocean Acidification

## Corals Are Dissolving Away

- ..researchers monitored 57 locations at five coral reefs around the world... [and] found a strong correlation between the dissolving process and calcium carbonate levels in the water...the dissolving process ... [is] sensitive to ocean acidification...
- By Chelsea Harvey & E&E News, February 23, 2018
- <https://www-scientificamerican-com.tudelft.idm.oclc.org/article/corals-are-dissolving-away1/>
- Bradley D. Eyre et al. ,Coral reefs will transition to net dissolving before end of century.Science **359**,908-911(2018).DOI:10.1126/science.aao1118
- <https://www-science-org.tudelft.idm.oclc.org/doi/10.1126/science.aao1118>



# Probability Calculations – a summary of the lecture

- First note that we are after  $P(D \cap S)$ , where  $P(D) = 0.2$  and  $P(S) = 0.4$ .
- $P(D \cap S) = P(S)P(D) = 0.08$  **iff** they are independent. This is not always a good assumption.
- Using empirical (data on next slide), we see 0.10 compared to 0.08.
- Challenge:  $\rightarrow$  how do we get  $P(D \cap S)$ ?
  - $\rightarrow$  a multivariate distribution!
- Regarding these distributions, we covered this in lecture:
  - How to integrate density in 2D to get a probability
  - What the contours are and what they tell us about probability
  - Correlation coefficient and how that influences contours/probability



## Data\*: Observations of Coral Atolls

D  $\rightarrow$  [2.1, 2.6, 4.3, 3.8, 2.5, 4.7, 1.4, 1.9, 3.6, 3.1]

S  $\rightarrow$  [5.1, 3.2, 7.2, 4.8, 6.5, 4.1, 2.4, 6.2, 6.9, 3.6]

Elevation of island is 10 m above sea level.

Find probability that it is underwater after 4 m settlement and 6 m sea level rise.

## Data\*: Observations of Coral Atolls

$D \rightarrow [2.1, 2.6, \mathbf{4.3}, 3.8, 2.5, 4.7, 1.4, 1.9, 3.6, 3.1]$

$S \rightarrow [5.1, 3.2, \mathbf{7.2}, 4.8, 6.5, 4.1, 2.4, 6.2, 6.9, 3.6]$

$$P[D \cap S] = \frac{1}{10} \rightarrow \neq P(A)P(S) \text{ and } \neq (1 - F_D(d = 4))(1 - F_S(s = 6))$$

Why?!

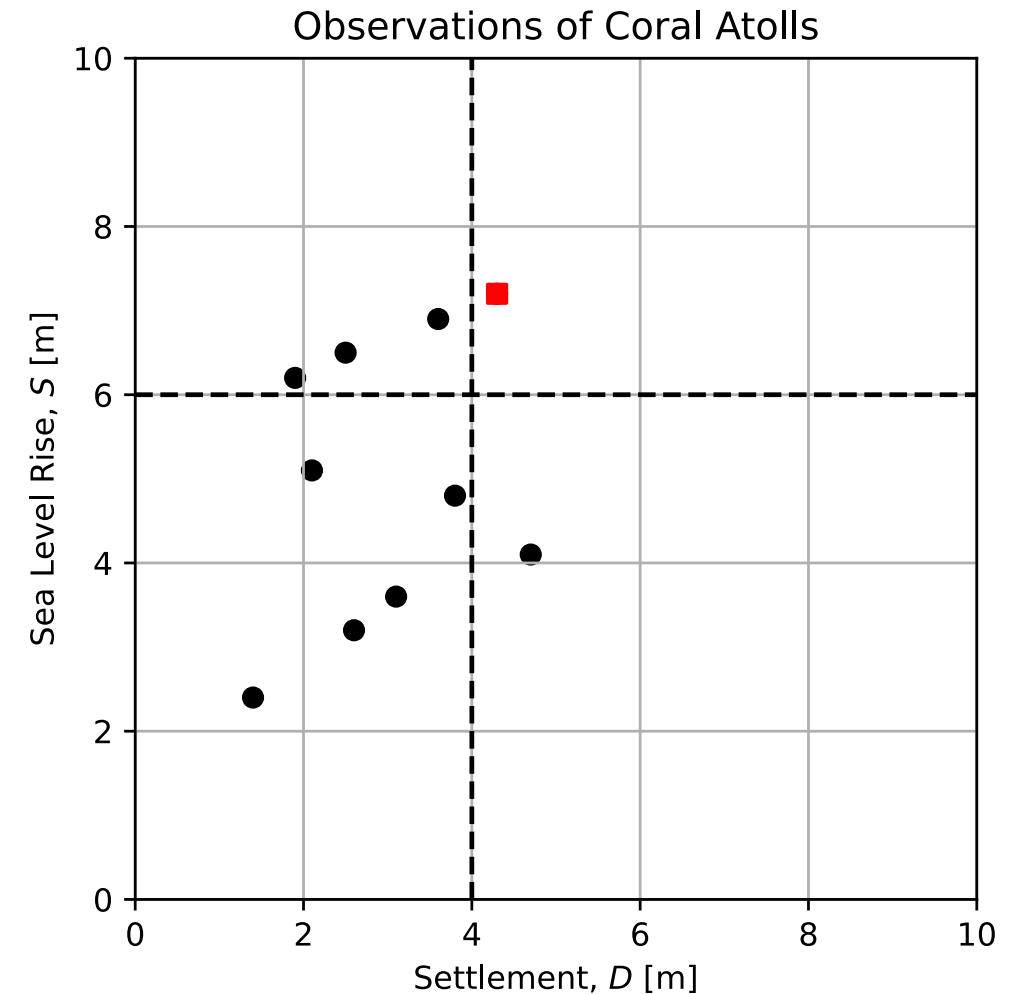
# Data\*: Observations of Coral Atolls

$D \rightarrow [2.1, 2.6, \mathbf{4.3}, 3.8, 2.5, 4]$

$S \rightarrow [5.1, 3.2, \mathbf{7.2}, 4.8, 6.5, 4]$

$$P[D \cap S] = \frac{1}{10} \rightarrow \neq P(A)P(S) \text{ and } \neq$$

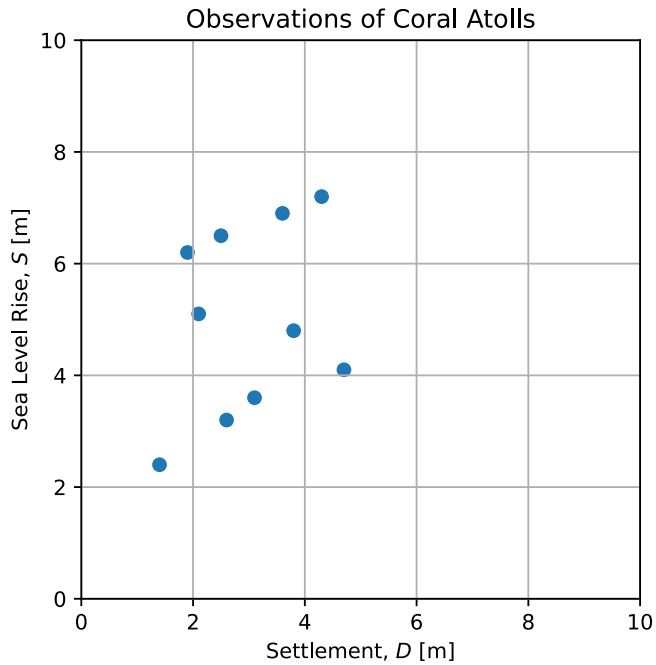
Why?!



# Covariance, Dependence

- Definitions (refresher)
- Influence of (Pearson) correlation coefficient,  $\rho$
- Bivariate perspective

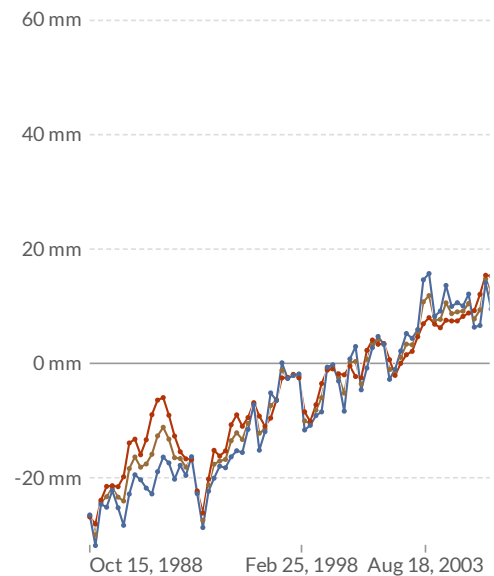
# Settlement v. Sea Level Rise?



Not covered in lecture.  
Illustration that dependence plays a role.

## Sea level rise

Global mean sea level rise is measured relative to a series: the widely-cited Church & White dataset; the average of the two.



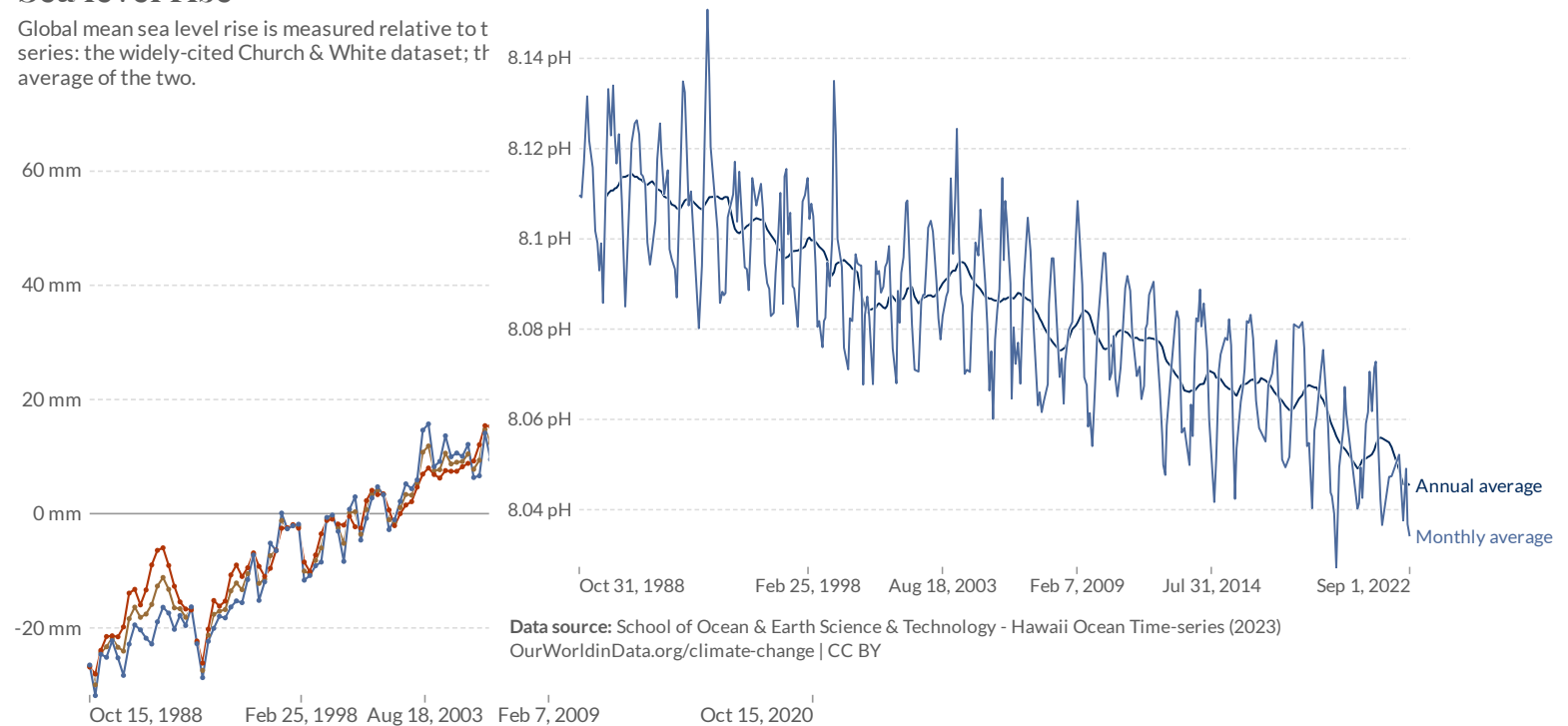
## Global atmospheric CO<sub>2</sub> concentration

Atmospheric carbon dioxide (CO<sub>2</sub>) concentration is measured in parts per million (ppm).



## Ocean acidification: mean seawater pH, Hawaii

Mean seawater pH is shown based on in-situ measurements of pH from the Aloha station in Hawaii.



Data source: School of Ocean & Earth Science & Technology - Hawaii Ocean Time-series (2023)  
OurWorldinData.org/climate-change | CC BY

## *D* and *S* as Random Variables (not events)

$$D \sim N(\mu = 3.0, \sigma = 1.0)$$

$$S \sim N(\mu = 5.0, \sigma = 1.6)$$

Elevation  $Z = 10$  m above sea level

## Function of random variables

- Define a function  $Z(D, S)$  to describe underwater case

$$Z(D, S) = 10 - D - S$$

Find probability  $P[Z < 0]$

# Function of random variables

$$Z(D, S) = 10 - D - S$$

$$D \sim N(\mu = 3.0, \sigma = 1.0) \quad S \sim N(\mu = 5.0, \sigma = 1.6)$$

$$\rho(X_1, X_2) = 0.3$$

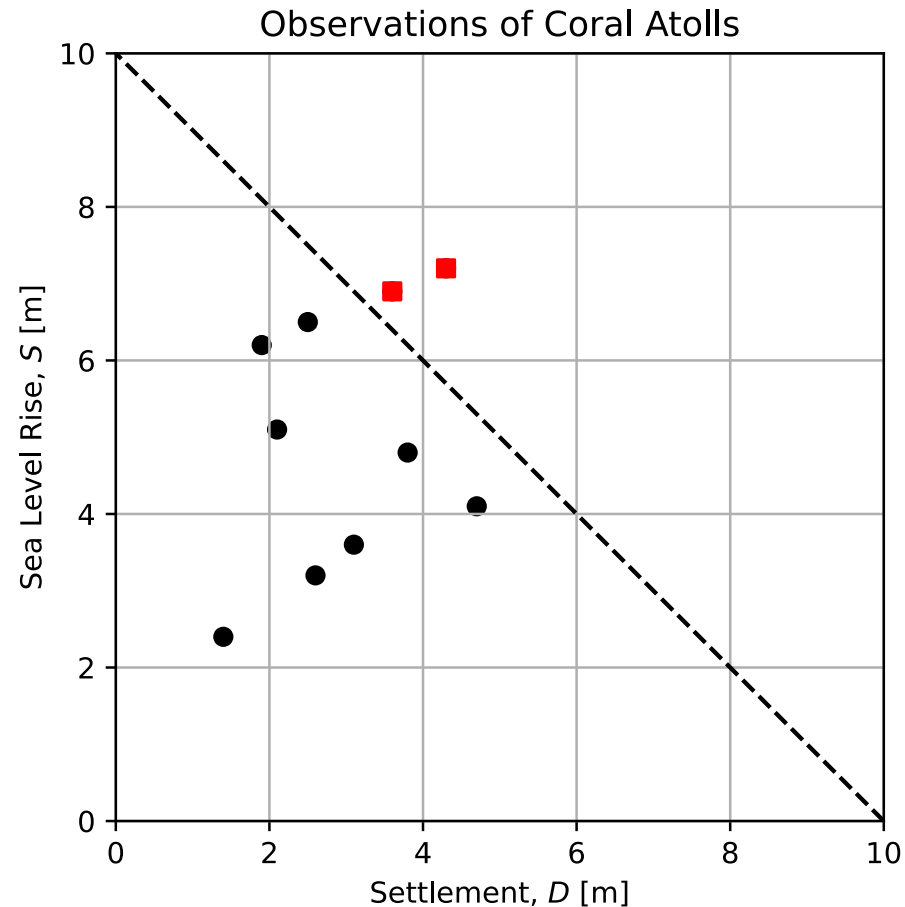
Not covered in lecture.  
Can evaluate analytically!

Find probability  $P[Z < 0]$



# Function of random variables

$$Z(D, S) = 10 - D - S$$



Not covered in lecture.  
Can evaluate analytically!

# Multivariate Distributions

- Independent Marginals
- Multivariate Gaussian
- Copulas

→ Density contours and samples in 2D are useful!

# Copula Page in Book

## 💡 Definition of bivariate copula

The definition of copula for the bivariate case is given by

$$F_{X_1, X_2}(x_1, x_2) = C[F_{X_1}(x_1), F_{X_2}(x_2)]$$

where  $F_{X_1, X_2}(x_1, x_2)$  for  $(x_1, x_2) \in \mathbb{R}^2$  is a joint distribution with marginals  $F_{X_1}(x_1)$  and  $F_{X_2}(x_2)$  in  $[0, 1]$  and  $C$  is a copula in the unit square  $I^2 = ([0, 1] \times [0, 1])$ , being this equation satisfied for all  $(x_1, x_2) \in \mathbb{R}^2$ .

Therefore, the joint density is given as the product of the density of the copula,  $c$ , and the densities of the marginals as

$$f_{X_1, X_2}(x_1, x_2) = f_{X_1}(x_1)f_{X_2}(x_2)c(F_{X_1}(x_1), F_{X_2}(x_2))$$

# Non-linearity

- Functions
- Marginal Distributions
- Dependence

# Programming Aspects

- Multivariate Distributions
- Monte Carlo Simulation

# Copula in Python (pyvinecopulib)

- `X = stats.multivariate_normal(mean=[3, 5],`
- `cov=[[1.00, rho],`
- `[rho, 2.60]])`

$$X = [X_1 \quad X_2 \quad \dots \quad X_m]^T$$

$$f_X(x) = \frac{1}{\sqrt{\det(2\pi\Sigma_X)}} \exp\left(-\frac{1}{2}(x - \mu_X)^T \Sigma_X^{-1}(x - \mu_X)\right)$$

# Copula in Python (pyvinecopulib)

- `X1: scipy.stats.rv_continuous` object. Marginal distribution
- `X2: scipy.stats.rv_continuous` object. Marginal distribution
- `rho: float`. Pearson correlation coefficient.
- `copula: cop.Bicop(family=pyvinecopulib.BicopFamily.gaussian, parameters = [rho])`

- `x1 = x[0]`
- `x2 = x[1]`
- `u =X1.cdf(x1)`
- `v =X2.cdf(x2)`
- `Return copula_uv.cdf(np.array([[u, v]]))[0]`

$$F_{X_1, X_2}(x_1, x_2) = C[F_{X_1}(x_1), F_{X_2}(x_2)]$$

$$f_{X_1, X_2}(x_1, x_2) = f_{X_1}(x_1)f_{X_2}(x_2)c(F_{X_1}(x_1), F_{X_2}(x_2))$$

# Monte Carlo Simulation

Associated terms: propagation of uncertainty, realizations, samples

Steps:

1. Define distributions for random variables (probability density over a domain)

Hint: `xxx.rvs(size=N)`

2. Generate random samples



A method of parent class **rv\_continuous**

3. Do something with the samples (deterministic calculation)

4. Evaluate the results: e.g., "empirical" PDF, CDF of samples, etc.

When RV's are *not* independent, make sure you sample from the joint distribution!

→ `scipy.stats.multivariate_normal.rvs()`



## Wednesday and Friday: build on last week, add...

- Bivariate joint distributions:
  - linear (Gaussian) dependence structure
  - non-Gaussian marginals
- Region of interest: thresholds, functions, etc.
- Contour plots
- Evaluating the effect of dependence (qualitative and quantitative)

Thanks for joining the...



Modelling, Uncertainty, and Data for Engineers

Last Q1  
Lecture!