

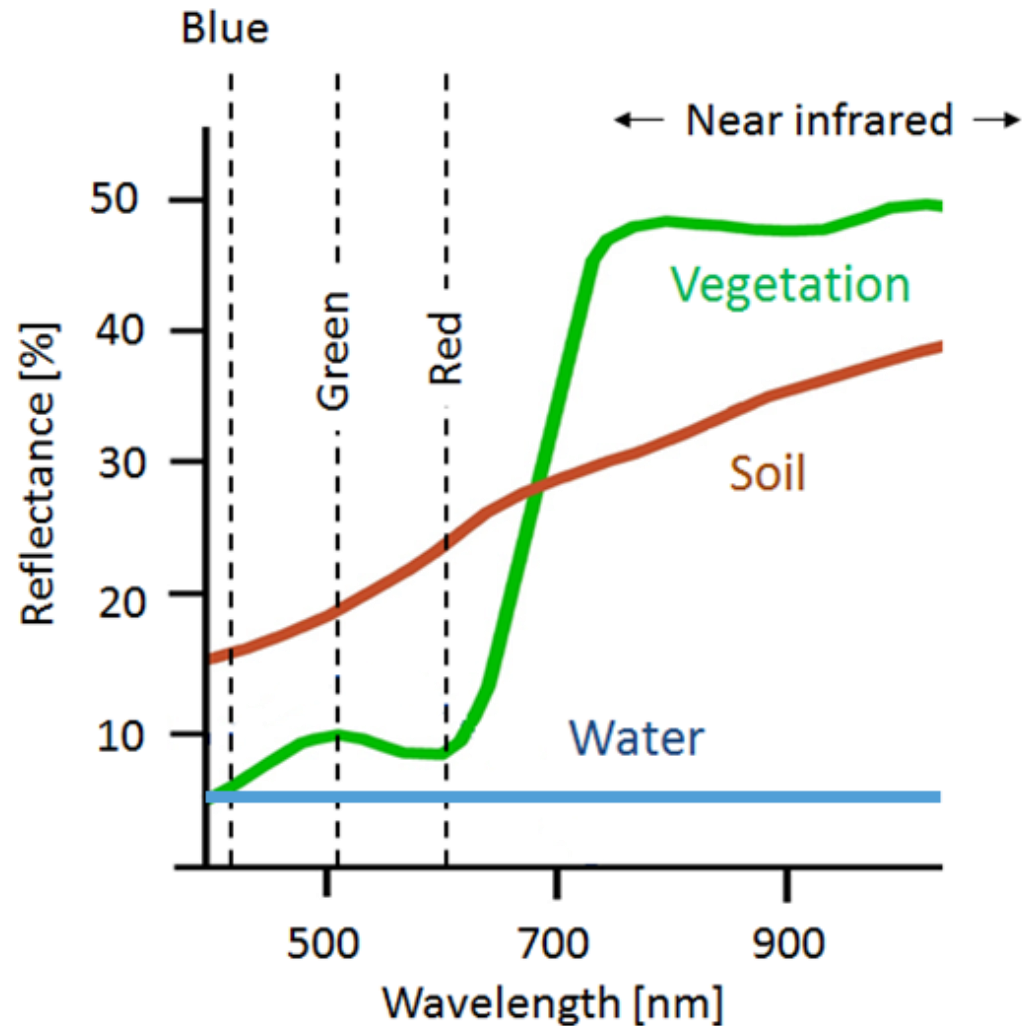
Vegetation Indices – 1st order Taylor

Uncertainty propagation in spatial environmental modelling

2024, Sytze de Bruin



Background: spectral signatures land cover



Vegetation indices

- Use the jump around the “red edge” to characterize vegetation
- There are many of them, e.g.:
 - $NDVI = (NIR - RED) / (NIR + RED)$
 - $SR = NIR / RED$
- They are frequently used in vegetation mapping
- Errors in the inputs propagate into the indices
- Demonstrated with case study
- Today: 1st order Taylor Series method

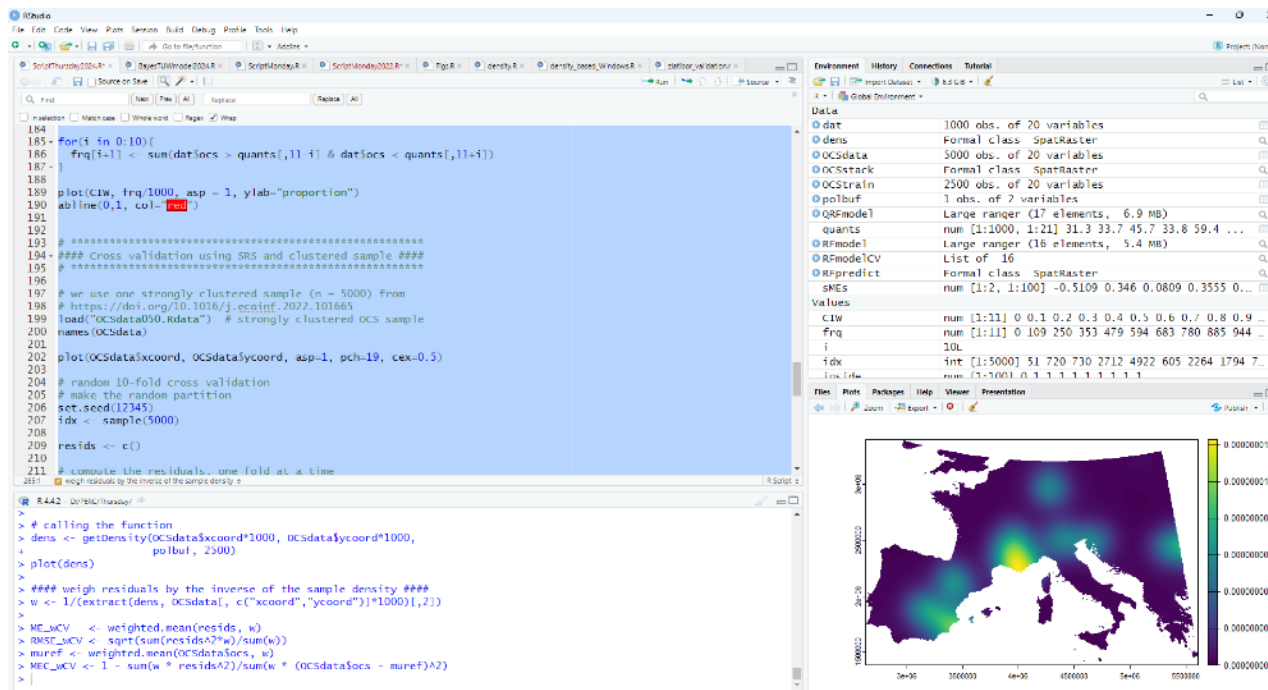
Microsoft Teams

The screenshot displays the Microsoft Teams application window. On the left sidebar, there are icons for Activity, Chat, Teams, Calendar, Calls, OneDrive, and Apps. The main area shows the 'General' channel of a team named 'US'. The breadcrumb path is 'Documents > General > 1 Monday > Practical'. Below this, a table lists files:

Name	Modified	Modified By
FCLorraine.zip	Tuesday at 10:35 AM	Bruin, Sytze de
VegIndicesMonday2024.pdf	Tuesday at 10:47 AM	Bruin, Sytze de

R and RStudio

- Make and save script file
- Complete script will be provided at end of the day



The practical / tutorial

- Instructions in file
 - Code block
 - Snippets
 - Complete code (Wednesday – Friday)
- Go through the tutorial & interpret results
- Questions are to help thinking about the matter
- At ~ 15:45 Feedback
- Ask us if something is unclear / does not work
- Any Mac users? → <https://www.xquartz.org/index.html>

1) Derivation Tau(SR)

From chapter 14: $\tau^2 \approx \sum_{i=1}^m \sum_{j=1}^m \rho_{ij} \sigma_i \sigma_j g'_i(\bar{b}) g'_j(\bar{b})$

SR = NIR/RED

$f'_{NIR} = 1/RED$ (multiplication by constant)

$f'_{RED} = -NIR/RED^2$ (power rule, if $f(x) = x^n$, then derivative $f'(x) = nx^{n-1}$)

Fill in: $\tau^2 \approx \sum_{i=1}^m \sum_{j=1}^m \rho_{ij} \sigma_i \sigma_j g'_i(\bar{b}) g'_j(\bar{b})$

$$\tau_{SR}^2 \approx \frac{\sigma_{NIR}^2}{RED^2} + \sigma_{RED}^2 \cdot \frac{NIR^2}{RED^4} - 2\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot \frac{NIR}{RED^3}$$

2) Derivation Tau(NDVI)

$$\text{NDVI} = (\text{NIR} - \text{RED}) / (\text{NIR} + \text{RED})$$

Quotient rule, if $f(x) = g(x)/h(x)$ and $h(x) \neq 0$, then the derivative

$$f'(x) = \frac{h(x) \cdot g'(x) - h'(x) \cdot g(x)}{(h(x))^2}$$

$$f'_{\text{NIR}} = \frac{1 \cdot (\text{NIR} + \text{RED}) - (\text{NIR} - \text{RED}) \cdot 1}{(\text{NIR} + \text{RED})^2} = \frac{2 \text{ RED}}{(\text{NIR} + \text{RED})^2}$$

$$f'_{\text{RED}} = \frac{-1 \cdot (\text{NIR} + \text{RED}) - (\text{NIR} - \text{RED}) \cdot 1}{(\text{NIR} + \text{RED})^2} = \frac{-2 \text{ NIR}}{(\text{NIR} + \text{RED})^2}$$

Fill in: $\tau^2 \approx \sum_{i=1}^m \sum_{j=1}^m \rho_{ij} \sigma_i \sigma_j g'_i(\bar{b}) g'_j(\bar{b})$

NDVI (continued)

$$\begin{aligned}\tau_{NDVI}^2 &\approx \sigma_{NIR}^2 \cdot \left(\frac{2 RED}{(NIR + RED)^2} \right)^2 + \sigma_{RED}^2 \cdot \left(\frac{-2 NIR}{(NIR + RED)^2} \right)^2 \\ &\quad + 2\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot \frac{-2 NIR}{(NIR + RED)^2} \cdot \frac{2 RED}{(NIR + RED)^2} \\ &= \sigma_{NIR}^2 \cdot \frac{4 RED^2}{(NIR + RED)^4} + \sigma_{RED}^2 \cdot \frac{4 NIR^2}{(NIR + RED)^4} \\ &\quad - 2\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot \frac{4 NIR \cdot RED}{(NIR + RED)^4} \\ &= \frac{4(\sigma_{NIR}^2 \cdot RED^2 + \sigma_{RED}^2 \cdot NIR^2) - 8\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot NIR \cdot RED}{(NIR + RED)^4}\end{aligned}$$

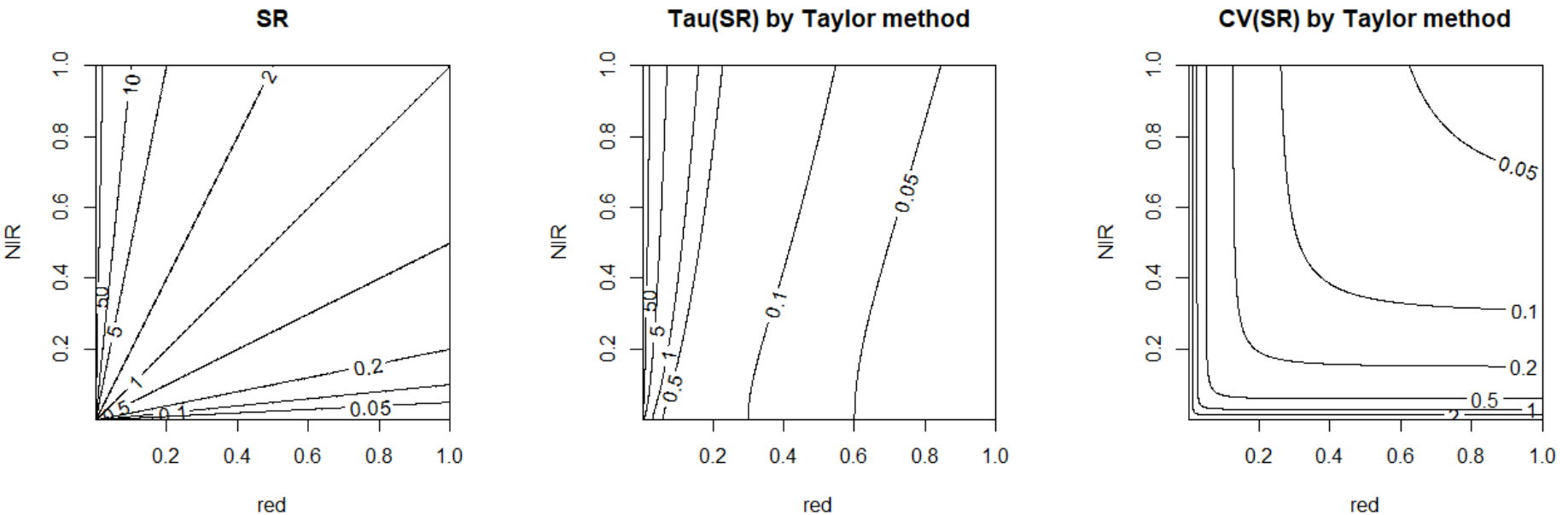
4) Function in R

```
# ===== Exercise 4 =====  
# Uncertainty NDVI by Taylor series method, returns tau.  
# Inputs as TaylorSR.  
TaylorNDVI <- function(red, nir, s_red, s_nir, rho) {  
  tau_sq <- 4*(s_nir^2*red^2+s_red^2*nir^2 -  
              2*rho*nir*red*s_red*s_nir)/(nir+red)^4  
  tau_sq[tau_sq < 0] <- 0  
  sqrt(tau_sq)  
}  
# =====
```

[Download from: MS Teams](#)

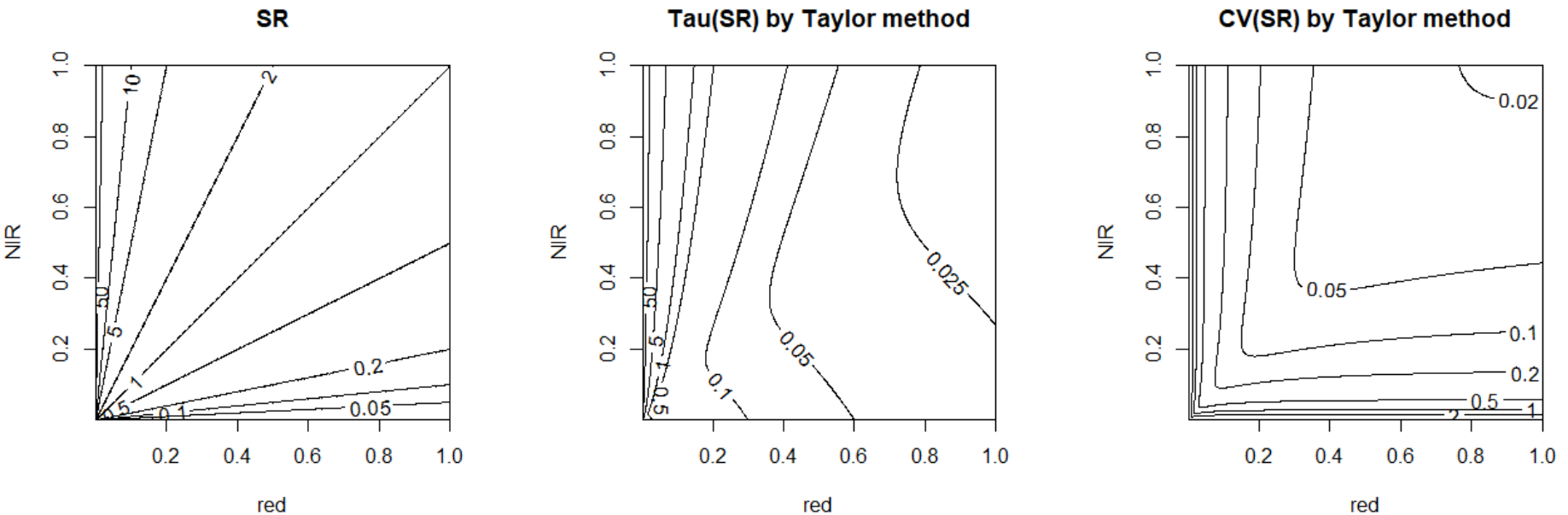
$$\frac{4(\sigma_{NIR}^2 \cdot RED^2 + \sigma_{RED}^2 \cdot NIR^2) - 8\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot NIR \cdot RED}{(NIR + RED)^4}$$

5) Plots τ_{SR} , 1st order Taylor, $\rho = 0$



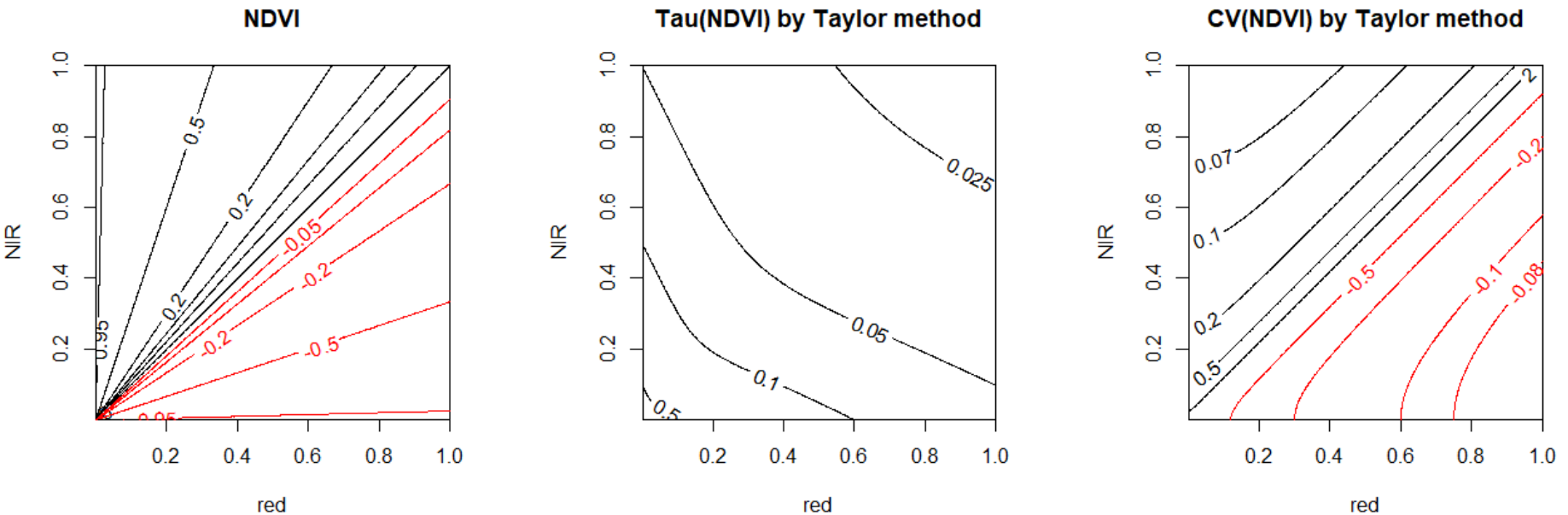
$$\tau_{SR}^2 \approx \frac{\sigma_{NIR}^2}{RED^2} + \sigma_{RED}^2 \cdot \frac{NIR^2}{RED^4} - 2\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot \frac{NIR}{RED^3}$$

5) Plots τ_{SR} , 1st order Taylor, $\rho = 0.8$



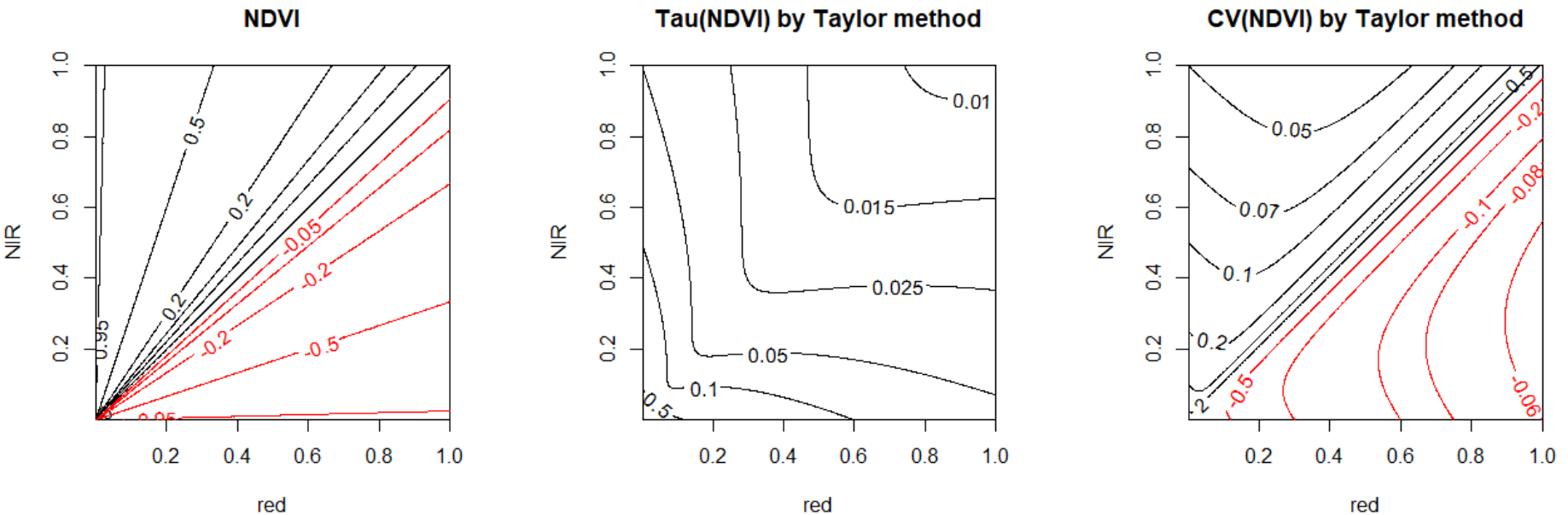
$$\tau_{SR}^2 \approx \frac{\sigma_{NIR}^2}{RED^2} + \sigma_{RED}^2 \cdot \frac{NIR^2}{RED^4} - 2\rho \cdot \sigma_{NIR} \cdot \sigma_{RED} \cdot \frac{NIR}{RED^3}$$

6) Plots τ_{NDVI} - 1st order Taylor, $\rho = 0$



$$\frac{4(\sigma_{\text{NIR}}^2 \cdot \text{RED}^2 + \sigma_{\text{RED}}^2 \cdot \text{NIR}^2) - 8\rho \cdot \sigma_{\text{NIR}} \cdot \sigma_{\text{RED}} \cdot \text{NIR} \cdot \text{RED}}{(\text{NIR} + \text{RED})^4}$$

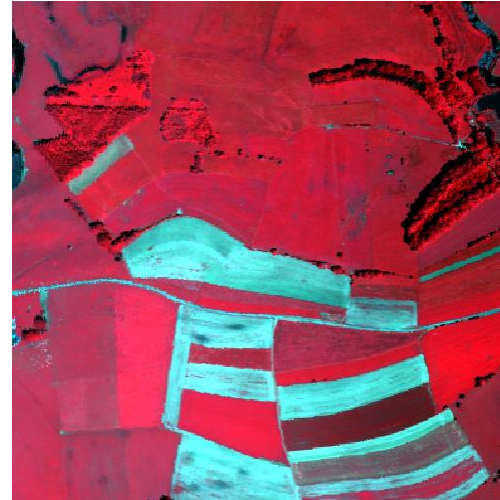
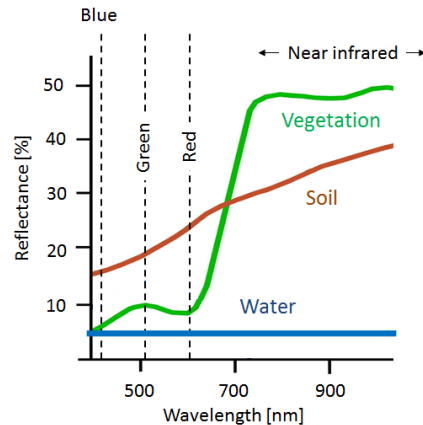
6) Plots τ_{NDVI} - 1st order Taylor, $\rho = 0.8$



$$\frac{4(\sigma_{\text{NIR}}^2 \cdot \text{RED}^2 + \sigma_{\text{RED}}^2 \cdot \text{NIR}^2) - 8\rho \cdot \sigma_{\text{NIR}} \cdot \sigma_{\text{RED}} \cdot \text{NIR} \cdot \text{RED}}{(\text{NIR} + \text{RED})^4}$$

7) Remotely sensed image

```
> plotRGB(false_color, 3, 2, 1, stretch='lin')
```



```
> summary(false_color, size=2e5)
```

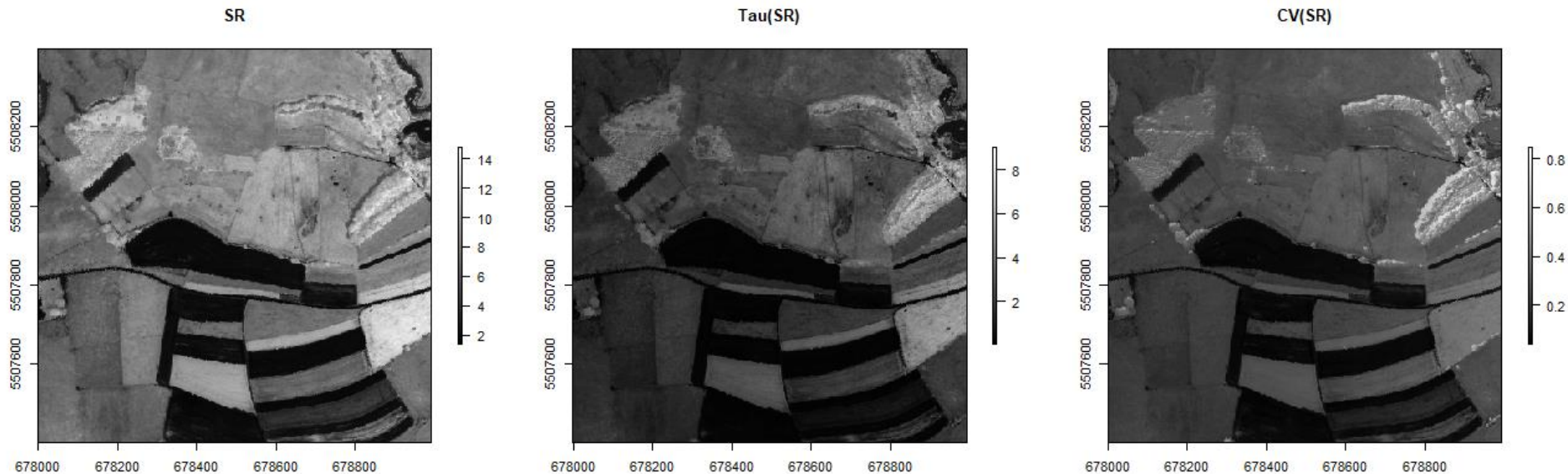
Layer_1		Layer_2		Layer_3	
Min.	:0.02136	Min.	:0.02699	Min.	:0.1069
1st Qu.:	0.08065	1st Qu.:	0.07160	1st Qu.:	0.5246
Median	:0.09480	Median	:0.08420	Median	:0.6137
Mean	:0.10701	Mean	:0.10806	Mean	:0.5911
3rd Qu.:	0.11412	3rd Qu.:	0.10425	3rd Qu.:	0.6579
Max.	:0.32644	Max.	:0.35923	Max.	:0.8906

8) Coefficient of variation SR

Apply a function to the cells of a
SpatRaster

```
CV_SR <- app(c(tau_SR, SRim), fun=function(x) {x[1]/x[2]},  
            filename="CV_SR.tif", overwrite=T)
```

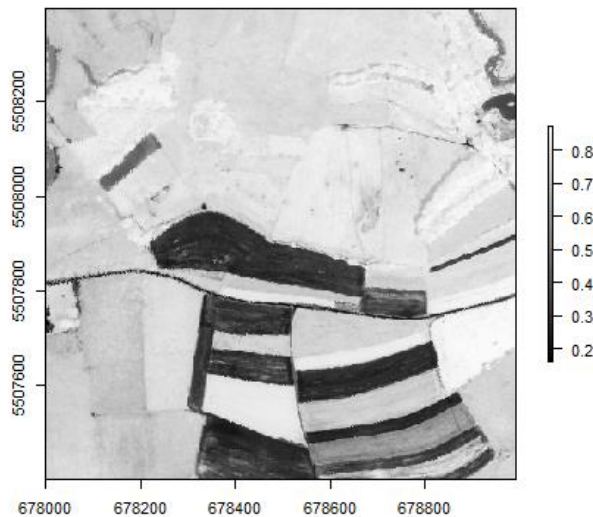
```
plot(CV_SR, col=gray(1:100/100), main= "CV(SR)", cex=1)
```



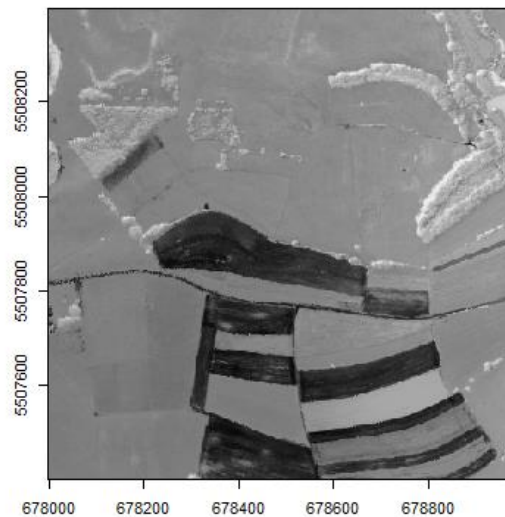
9) NDVI, Tau_NDVI and CV_NDVI

```
NDVIim <- app(false_color, fun=function(x) (x[3]-x[2])/(x[3]+x[2]),  
               filename="NDVI.tif", overwrite = T)  
plot(NDVIim, col=gray(1:100/100), main="NDVI")  
  
tau_NDVI <- app(false_color, fun=function(x)  
  TaylorNDVI(x[2], x[3], 0.025, 0.03, 0.8),  
  filename="tau_NDVI.tif", overwrite = T)
```

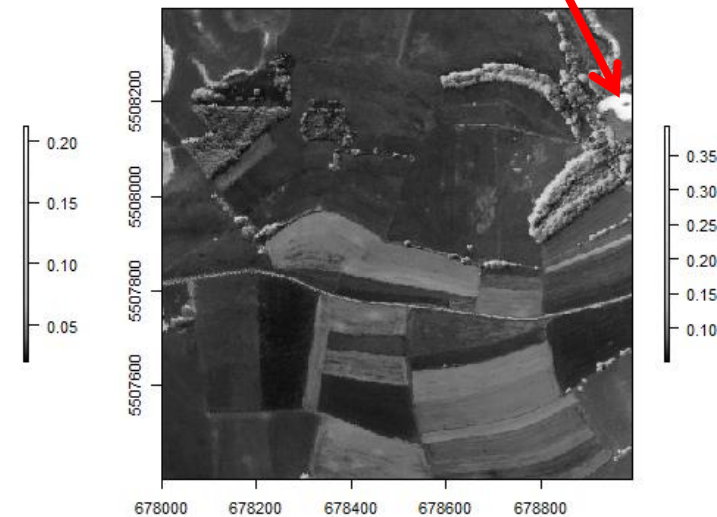
NDVI



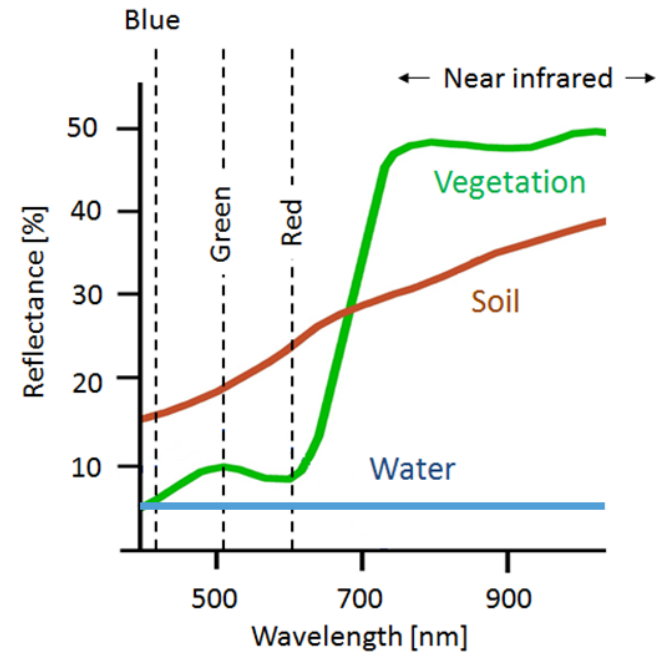
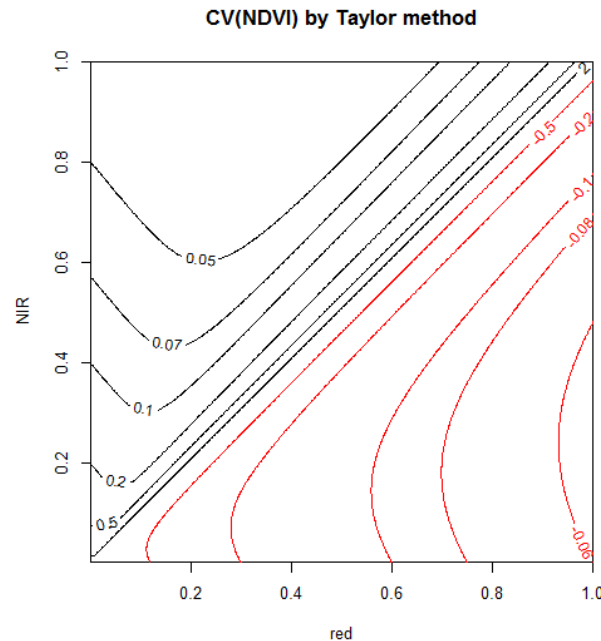
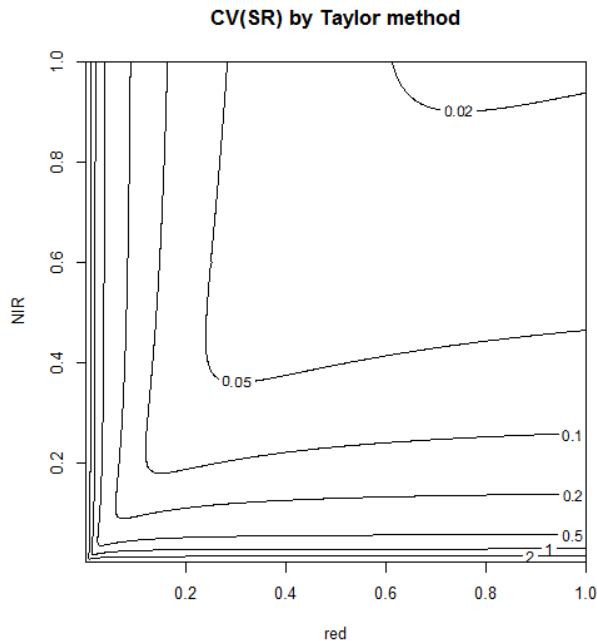
Tau(NDVI)



CV(NDVI)



The bright white spot in previous slide



Must be water!

<https://www.google.nl/maps/@49.6987691,5.4824111,515m/data=!3m1!1e3>

Take home

- 1st order Taylor series approximation gives a sum of terms, which can be easily interpreted
- Requires partial differentiation with respect to each considered uncertain input
- Introduces an approximation error in case of non-linear models
- Can be applied on spatial data
- It is fast!

Course dinner

- H41, Herenstraat 41
- 18:30

