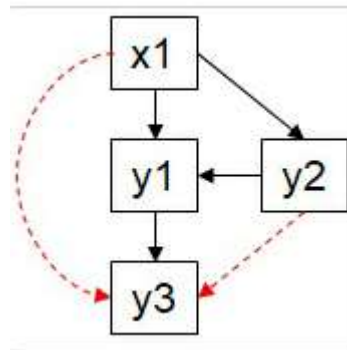


The advantages of piecewiseSEM

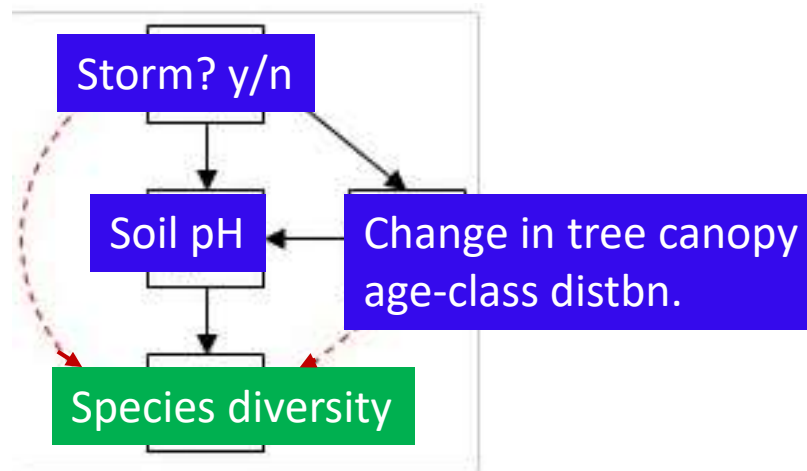
- Sit down quietly and construct a model where relationships are in series as well as in parallel and where these hypotheses reflect mechanisms and evidence at hand: black arrows below.



- The fit of your model is assessed against the missing paths: **red arrows**. In the jargon these are called ‘independence claims’.
- Previous methods did this by evaluating the whole model in one go (global estimation) but piecewiseSEM allows you to evaluate each model (local estimation) and then combine them. This is **REALLY** useful.

Applying expert knowledge before you start is critical...why?

- Because we are scientists!
- Avoids data-mining.
- This is not just a philosophical nicety. With only 4 variables there are 24 independence claims to test! (Shipley 2009. *Ecology* **90**, 363-368)
- Also *piecewiseSEM* will test any claims not hypothesised and so will tell you if you've missed any significant relationships.



Smart, et al (2014) *J.Ecol.* **102**, 1273-1287.

Explaining large-scale variation in a key soil quality indicator; Soil Aggregate Stability

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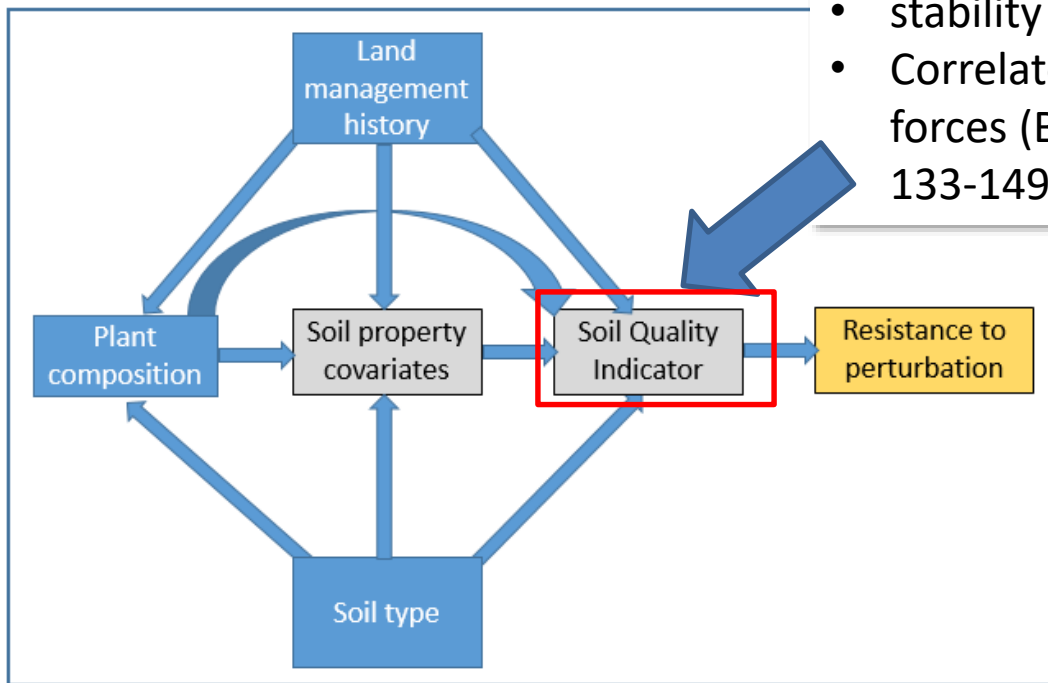
OBJECTIVE

Identify key controls on soil quality indicators.

Why? So we can manage for greater soil resilience given demand for food, conservation of soil C, climate change, extreme weather, changes in crops and land-use

Soil Aggregate Stability

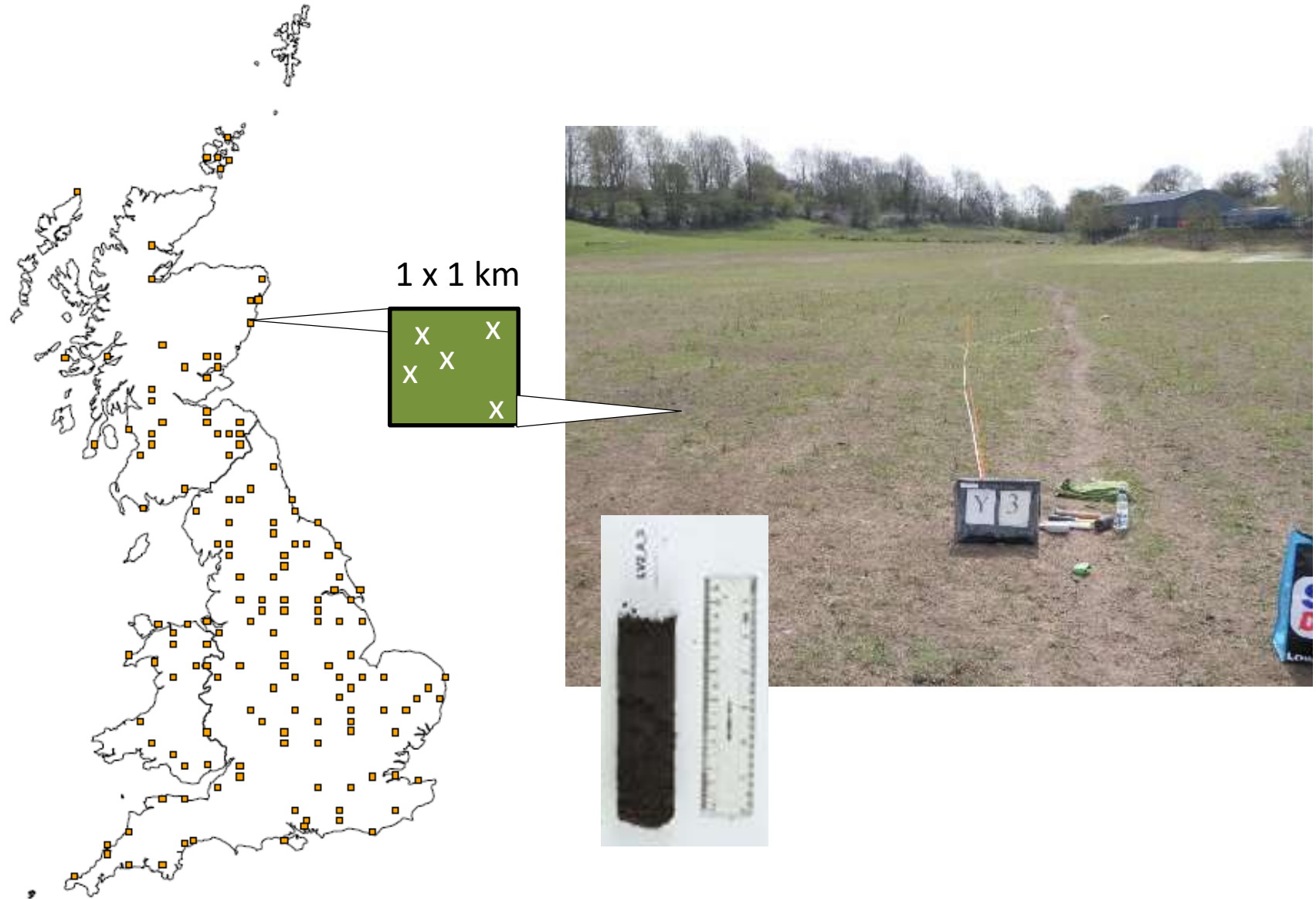
- stability of soil particles (<2mm diameter)
- Correlated with resistance to erosive forces (Barthès & Roose 2002 *Catena* **47**, 133-149)



“for every person on the planet an average 5 tonnes of soil are eroded every year” Quinton et al (2010) *Nature Geosci.* **3**, 311-314.

NATIONALLY REPRESENTATIVE SAMPLING DESIGN

#Vegetation (2x2m quadrat) and paired soil cores (5cm wide x 15cm deep) extracted from grassland and arable sampled by Countryside Survey in 2007 across Britain (n=286)



PLANT COMMUNITY & TRAIT-BASED PREDICTORS (2X2M QUADRAT):

- **Grass:forb ratio**
- **Plant species richness**
- **Above-ground NPP** (g dry mass/m²) – estimated from cover-weighted Leaf Dry Matter Content [1]; indicator of land-use intensity
- **Ordination (DCA) scores for vegetation axis 1**; strongly correlated with macronutrient availability [2,3]; indicator of land-use intensity
- **Specific Root Length** for dominant plant species extracted from FRED2.0 (m/g) [4]

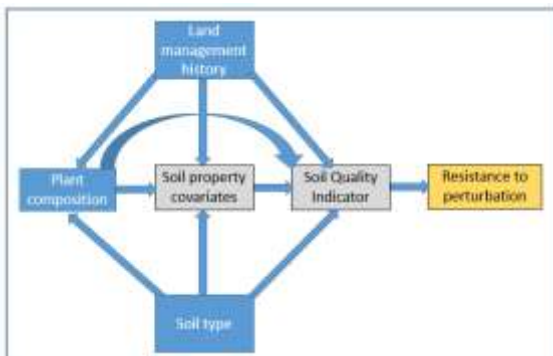
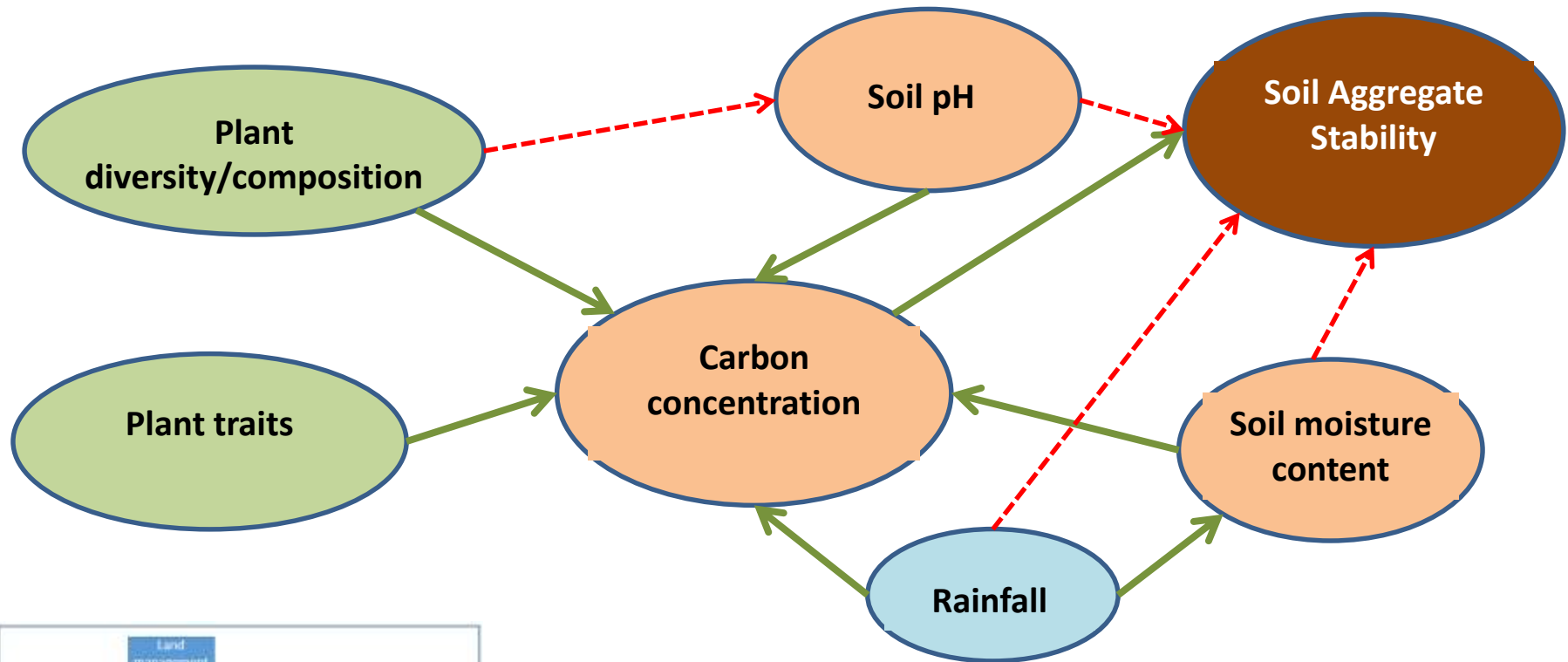
SOIL-DERIVED PREDICTORS (5X15CM CORE)[5]:

- **Carbon concentration** (g C/kg)
- **Volumetric soil moisture** (%)
- **Bulk density** (g/cm³)
- **Soil pH** (fresh soil)

1. Smart et al 2017. *Funct.Ecol.* **31**, 1336-1344.
2. Rowe et al 2014. *Environ.Sci.Processes Impacts*. doi: 10.1039/c4em00312hHjh
3. Smart et al 2003. *J.Env.Man.* **67**, 239-254.
4. Iversen et al. 2018. *FRED v.2* <https://doi.org/10.25581/ornlsfa.012/1417481>
5. Emmett et al. 2007. *Countryside Survey. Soils Report from 2007.*
www.countryside-survey.org.uk

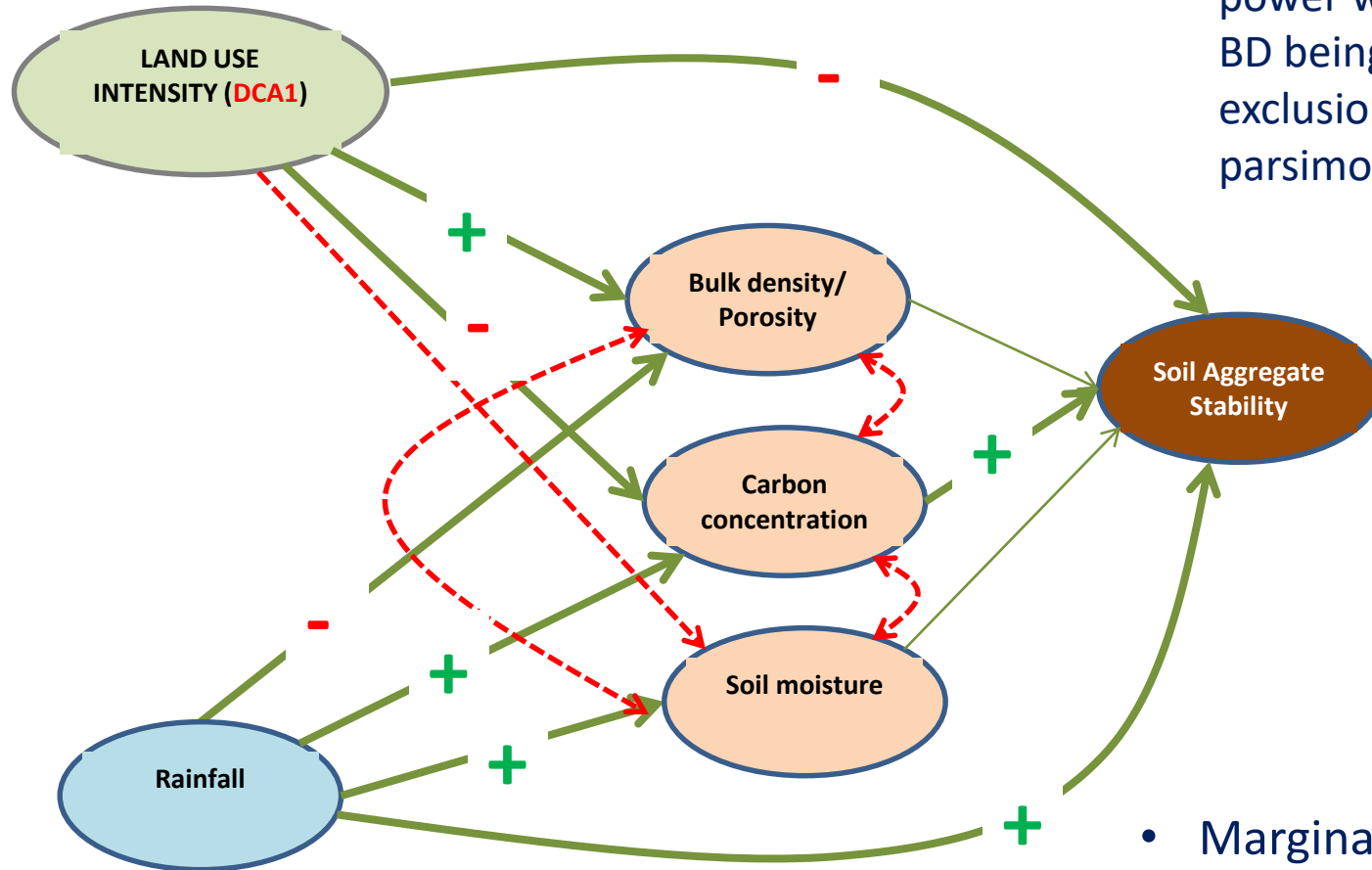


Plant traits, species composition, land-use intensity, soil pH, soil moisture and climate significant predictors of C fractions [1,2,3,4]. ... many links are missing though!



1. Wardle, DA et al 2012. *J.Ecol.* **100**, 16-30.
2. Manning, P et al 2015. *J.Appl.Ecol.* **52**, 1188-1196.
3. Ward, SE et al 2016. *Glob.Change.Biol.* **22**, 2929–2938.
4. Laliberté, E et al. 2012. *Ecology* **93**, 144-155.

#Model 1: AIC = 480, $P = 0$

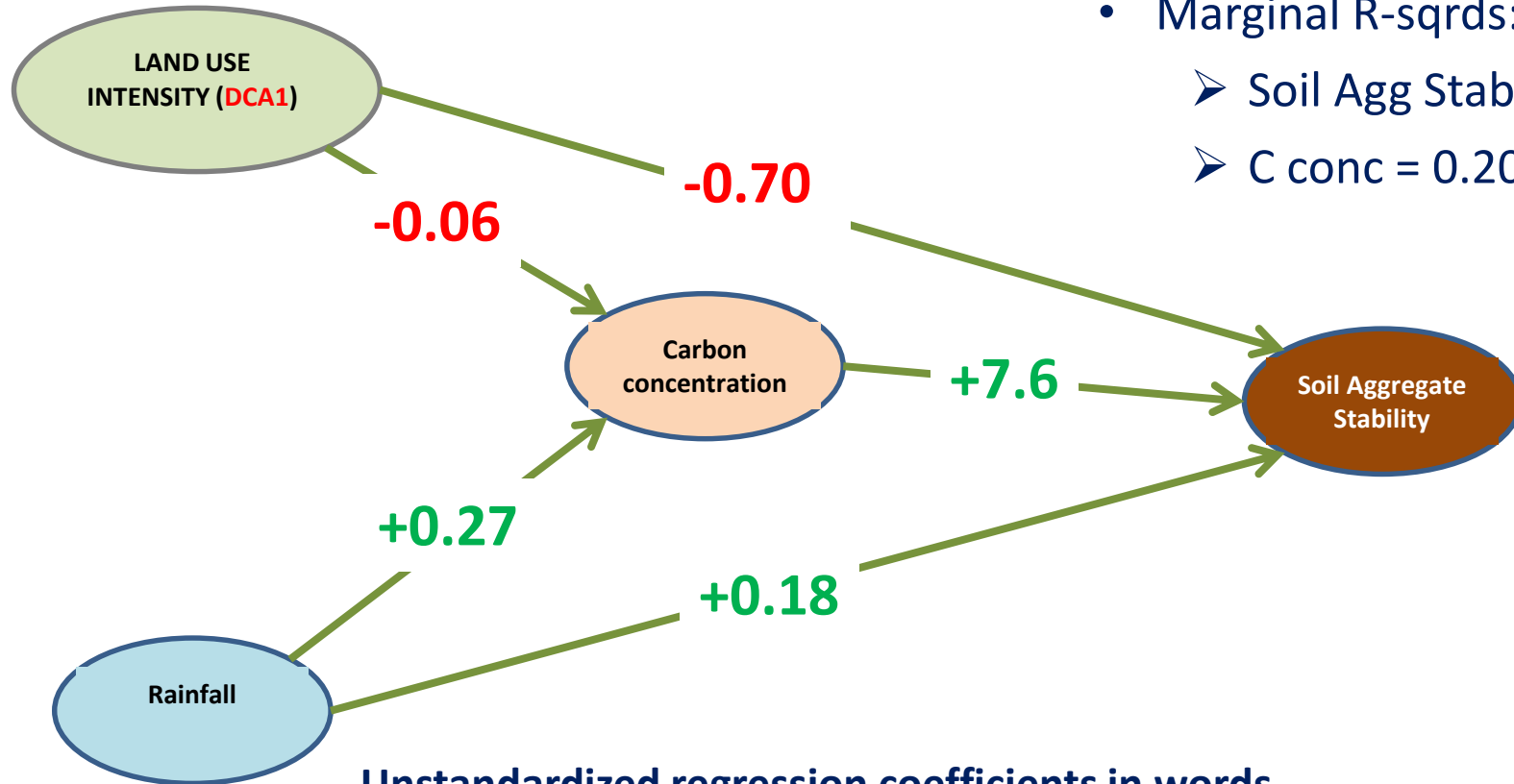


- **Basis set** includes highly correlated soil variables.
- They differ in explanatory power with soil moisture then BD being candidates for exclusion on grounds of parsimony.

- Marginal R-sqrds:
 - Soil Agg Stability = 0.55
 - C conc = 0.20

“Blindly selecting all the significant tests and re-inserting those paths, however, is irresponsible and in fact antithetical to the philosophy of SEM (*where paths are carefully specified by the user*).” J.Lefchek

#Model 4: AIC = 22. The best model.

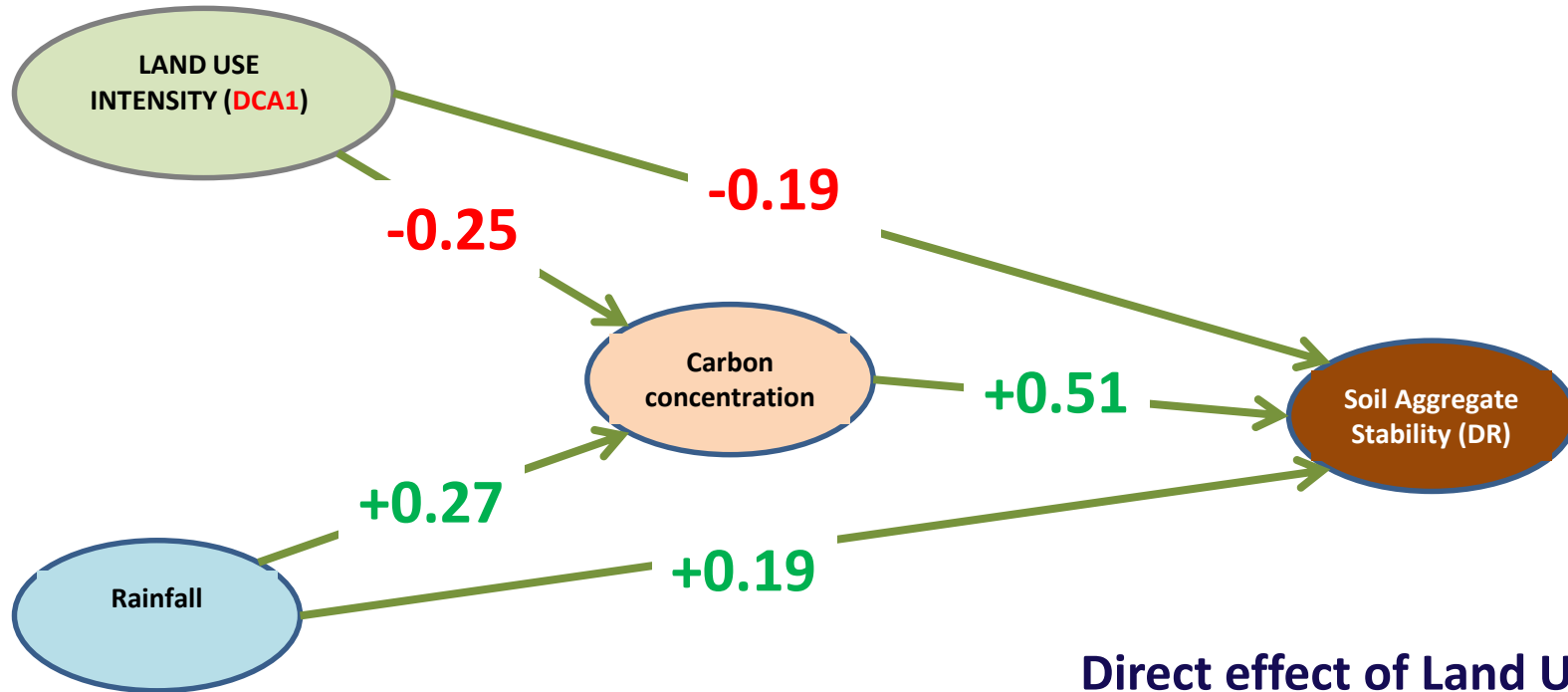


- It's simple at least!
- Marginal R-sqrds:
 - Soil Agg Stability = 0.54
 - C conc = 0.20

Unstandardized regression coefficients in words....

- An increase of 1 g C/kg **increases** Soil Agg Stability by 7.6 μm .
- An increased 10 mm of rainfall **increases** Soil Agg Stability by 1.8 μm .
- A 1% increase in land-use intensity **decreases** C concentration by 0.24 g C/kg and directly **decreases** Soil Agg Stability by 2.8 μm .

Getting more information from the output: **effect sizes** (mean / sdev) What would **es=1** mean?



Cohen (1988) suggested that $d=0.2$ be considered a 'small' **effect size**, 0.5 represents a 'medium' **effect size** and 0.8 a 'large' **effect size**. This **means** that if two groups' **means** don't differ by 0.2 standard deviations or more, the difference is trivial, even if it is statistically significant.

Direct effect of Land Use on DR?

-0.19

Indirect effect of Land Use on DR?

$-0.25 * 0.51 = -0.13$

Total effect of Land Use on DR?

$(-0.25 * 0.51) + -0.19 = -0.318$

Sources of information on *piecewiseSEM*

- <https://github.com/jslefche/piecewiseSEM>
- https://jslefche.github.io/sem_book/index.html
- piecewiseSEM R package (Lefchek 2016 *Methods in Ecol. Evol.* **7**, 573-579)
- Jon Lefchek answers email queries especially since the package is evolving and he's keen to improve it/hear about issues.
- Check out the package *blavaan* though. Bayesian SEM combining *lavaan* and *Stan* or *JAGS*.