

Sequestering carbon in regeneration: An elaborate hoax or landholders making a difference?

Associate Professor Cris Brack (Honorary)
Cris.Brack@anu.edu.au

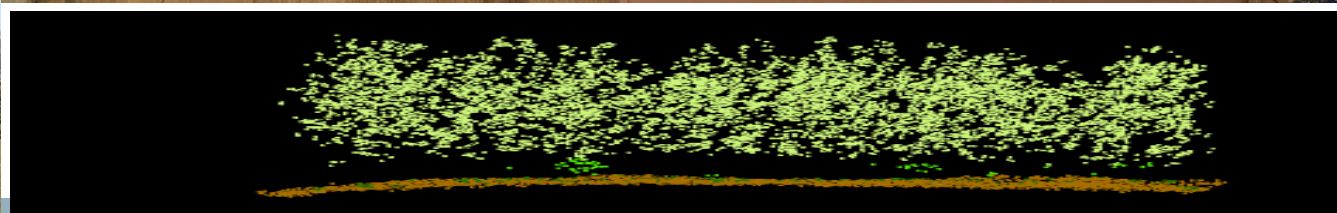
Human Induced Regeneration: to generate forests and greater carbon density





Australian
National
University

Lecturer in charge of toys



- All models are wrong...
 - some are useful...
 - in the right context
- All measurements are a compromise of:
 - Resources/time/money
 - Available tools/technology
 - Questions needing answers/decision points
 - Object itself



Object: What was/is/will be “forest”

May be classified as Forest if
sufficient plants to exceed 20% crown
cover in time

Classified as Forest (if at least 0.2 ha)



May be classified as Forest if
sufficient plants to exceed 20% crown
cover in time

Classified as Forest (if at least 0.2 ha)

NON-WOODY

(0-5% crown
cover)

SPARSE-WOODY

(>5-20% crown cover)

WOODLAND FOREST

(20-50% crown cover)

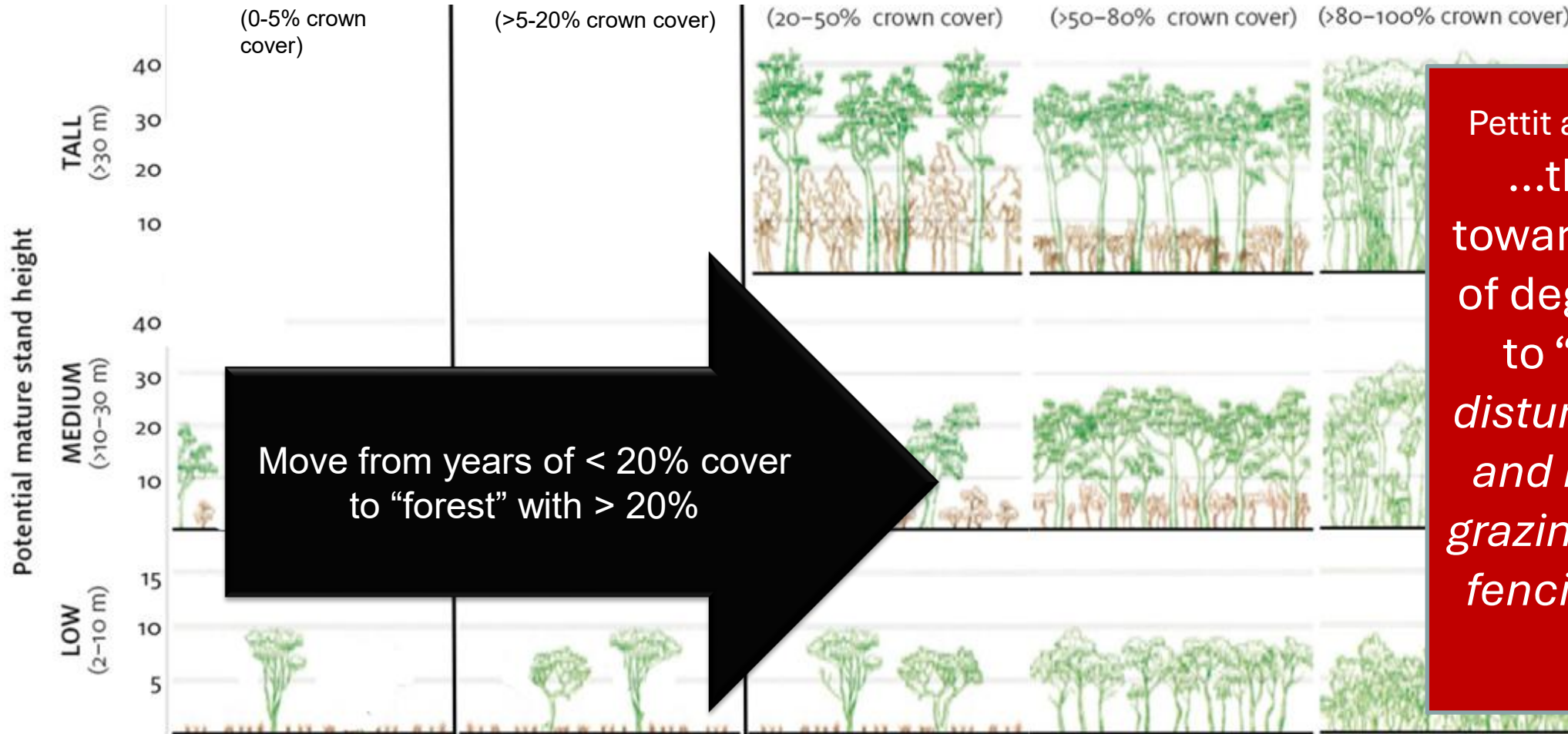
OPEN FOREST

(>50-80% crown cover)

CLOSED FOREST

(>80-100% crown cover)

WOODY



Pettit and Froend (2000)
...the first step
towards restoration
of degraded land is
to "remove the
disturbing element,
and in the case of
grazing this requires
fencing to exclude
stock..."

Criticism #1: no “permanence”

May be classified as Forest if
sufficient plants to exceed 20% crown
cover in time

Classified as Forest (if at least 0.2 ha)

NON-WOODY

(0-5% crown
cover)

SPARSE-WOODY

(>5-20% crown cover)

WOODLAND FOREST

(20-50% crown cover)

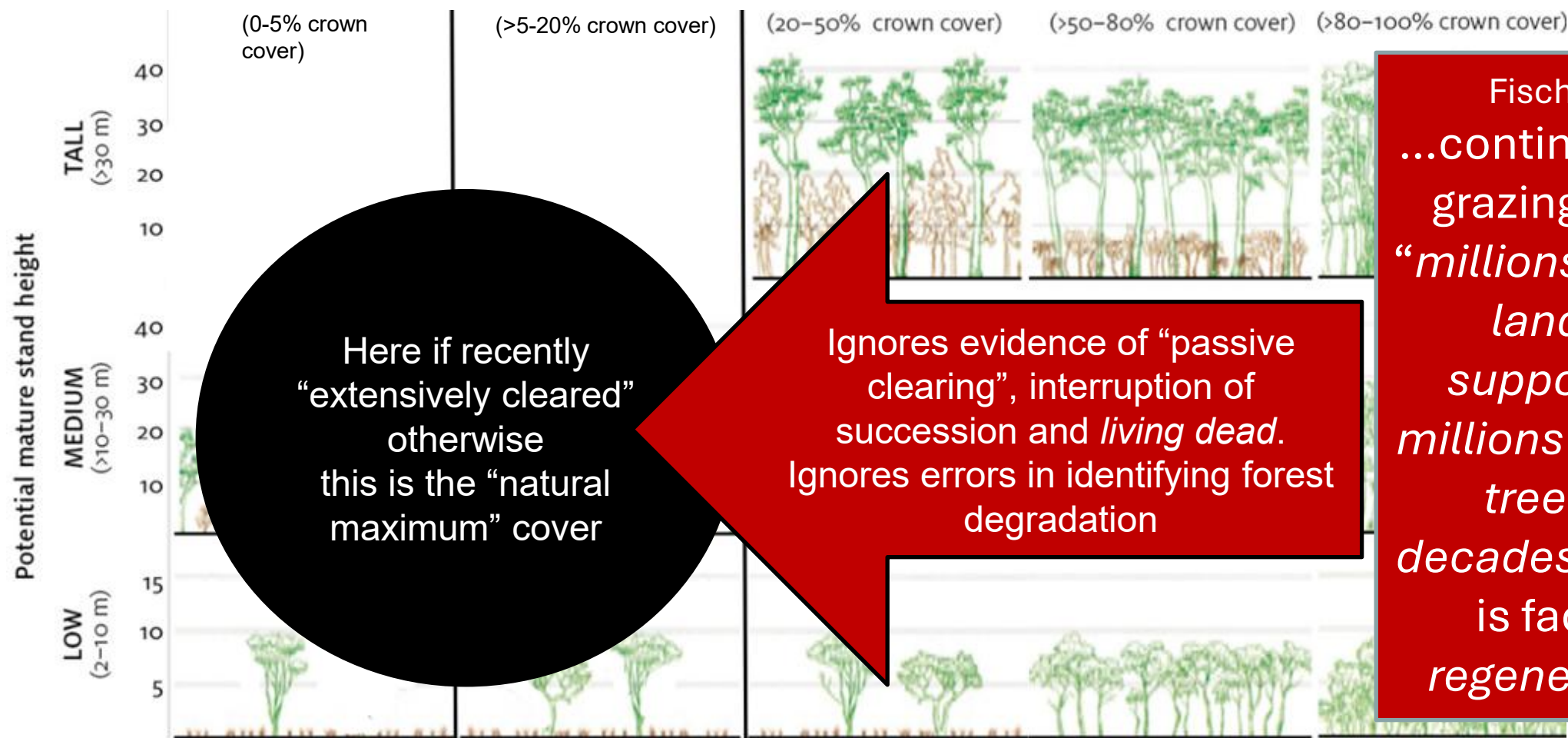
OPEN FOREST

(>50-80% crown cover)

CLOSED FOREST

(>80-100% crown cover)

WOODY



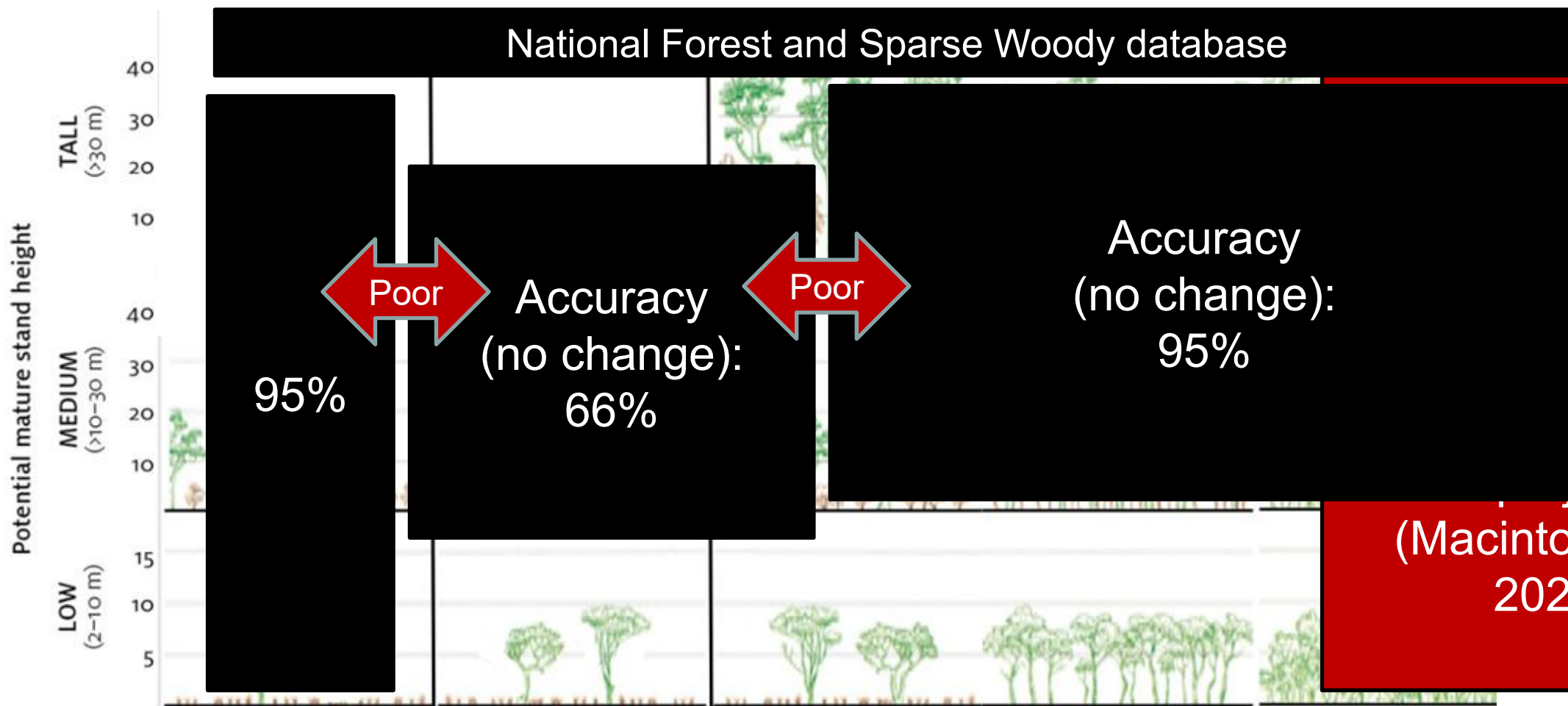
Fischer et al (2009)
...continuous livestock
grazing will result in
“*millions of hectares of
land currently
supporting tens of
millions of trees will be
treeless within
decades*” as the region
is facing “*a tree
regeneration crisis*”.

Measurement compromises: instruments

NON-WOODY

SPARSE-WOODY

WOODY



Assertion:
difference
R projects
hoc or
ic sample
ound the
ect
(Macintosh et al,
2024a)

Table 1 | Confusion matrix comparing classification results between NFSWVD (rows) and reference lidar dataset (columns) for 500 validation points, with associated omission and commission errors and overall accuracy (OA)

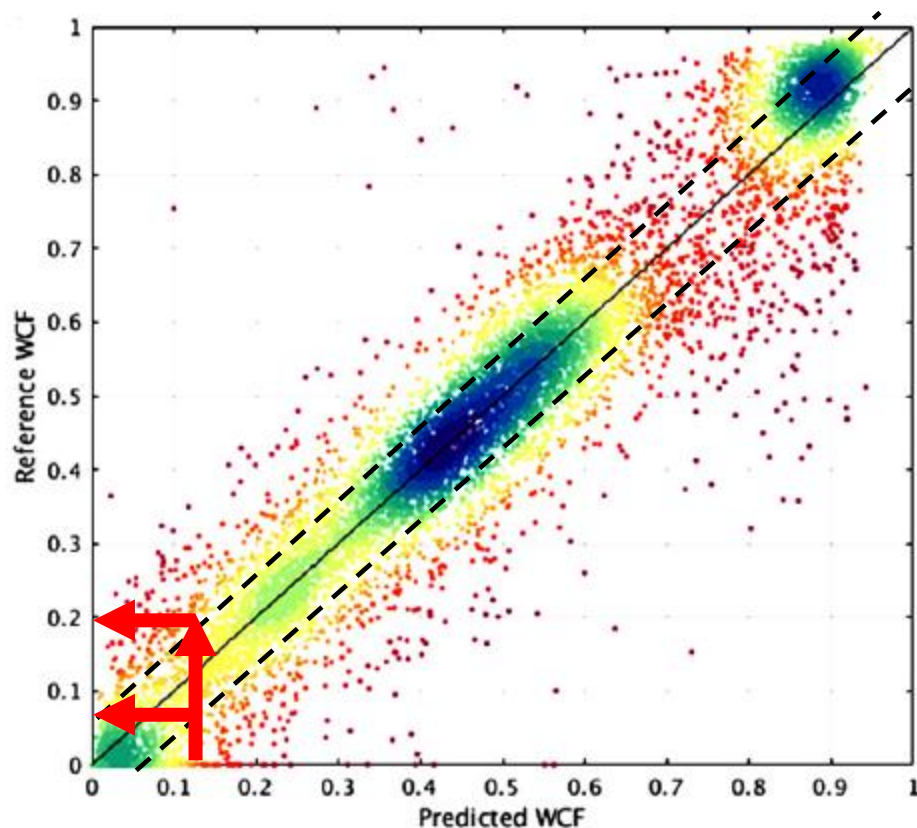
		Reference lidar				
		Non-woody	Sparse woody	Forest	Total	Commission
NFSWVD	Non-woody	131	97	42	270	51.5%
	Sparse woody	4	22	55	81	72.8%
	Forest	6	14	129	149	13.4%
	Total	141	133	226	500	
	Omission	7.1%	83.5%	42.9%		OA: 56.4%

Moore et al (2025)

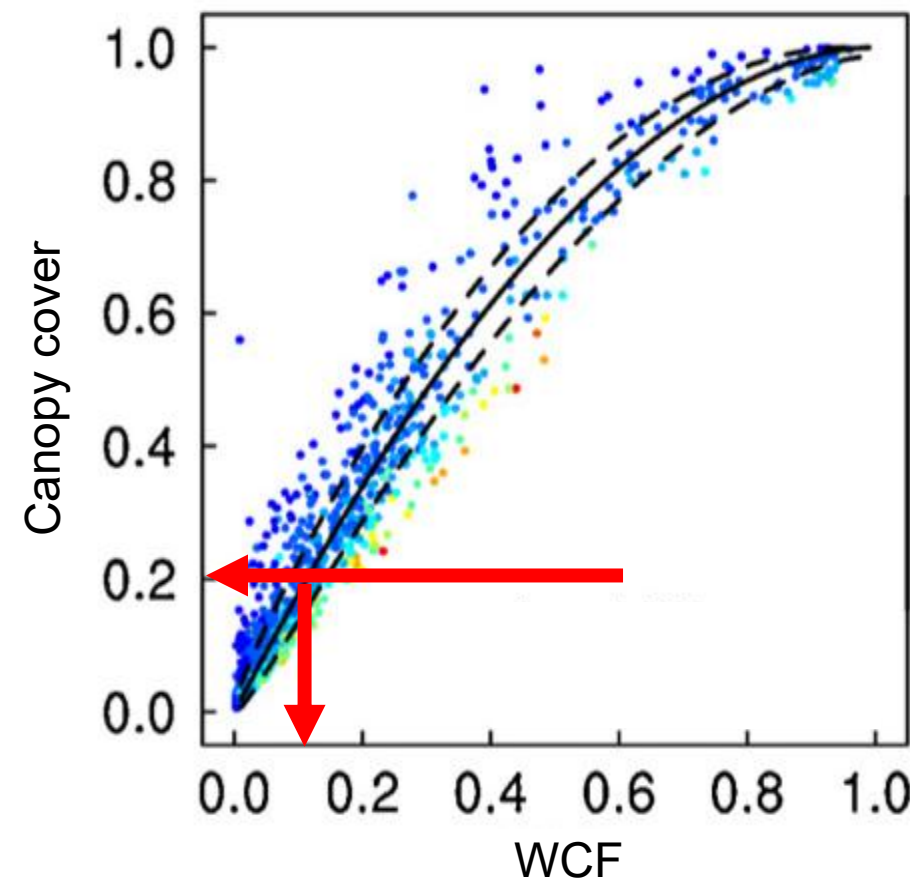
Is another national model useful?

Woody Cover Fraction (Liao et al, 2020) RMSE=0.07

Woody Cover Fraction -> Canopy Cover (Fischer et al, 2018) RMSE=0.07



Combined RMSE
 $= \sqrt{0.07^2 + 0.07^2}$
 ≈ 0.1



Such a RSME inflates type I error and reduces power of test ($H_0: \bar{u}_a = \bar{u}_b$ not able to reject)

Does NFSW support WCF?

Macintosh et al (2024b) undertook WCF analysis of canopy in HIR projects to provide *a way of cross-checking the results from Macintosh et al (2024a)* that relied on NFSW to assert that there was no difference between HIR and ad hoc samples

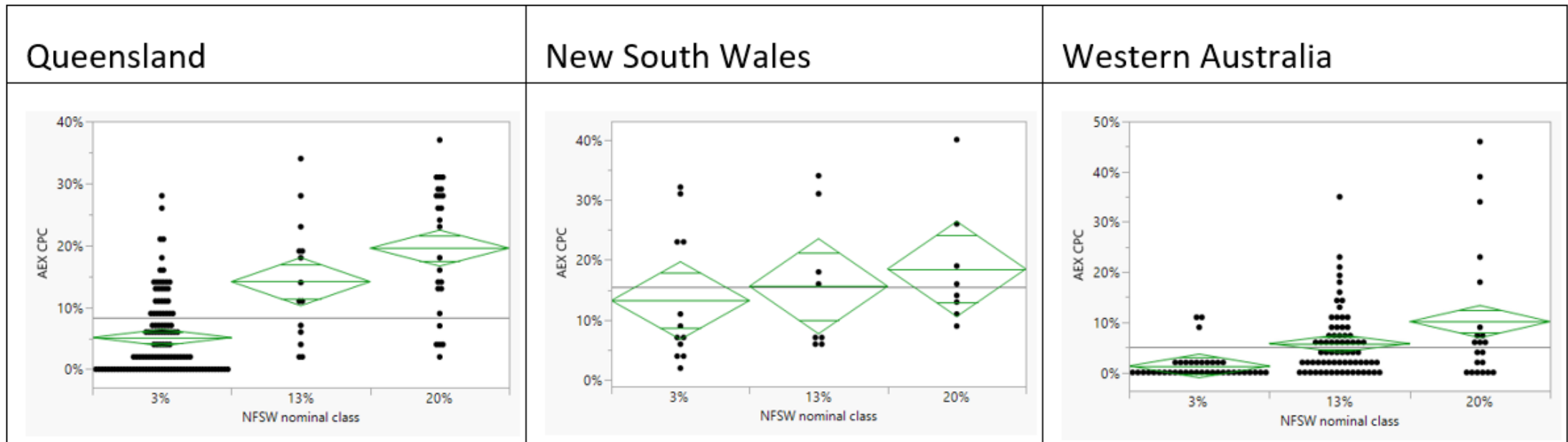
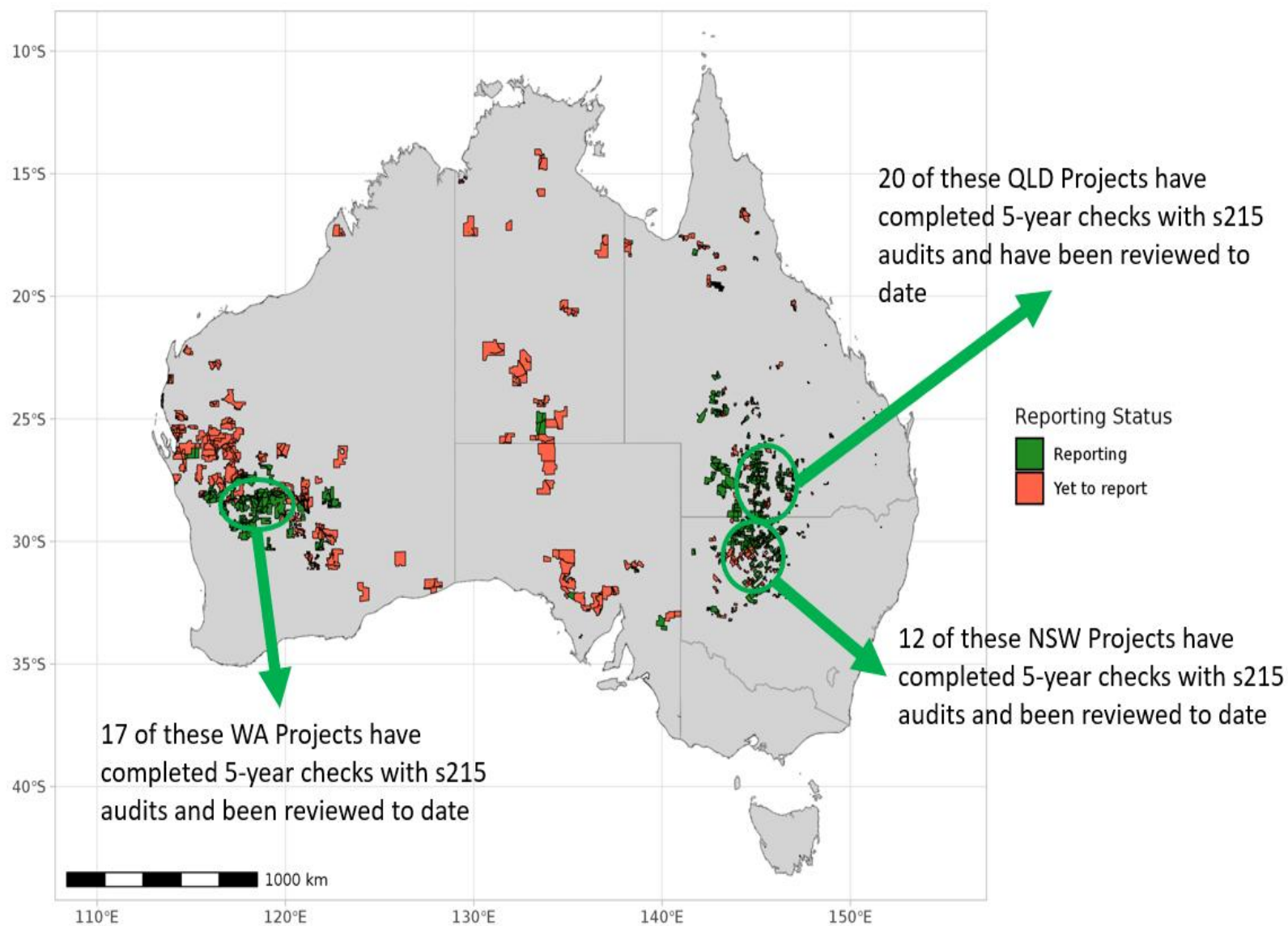


Figure 4: ANOVA of canopy cover estimated from AEX by NFSW classes (nominal values for non-woody, sparse-woody and woody respectively) for each State.

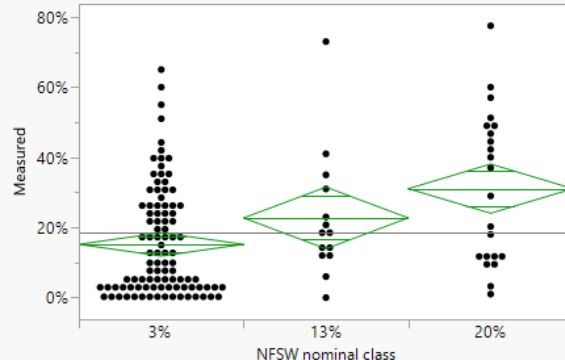
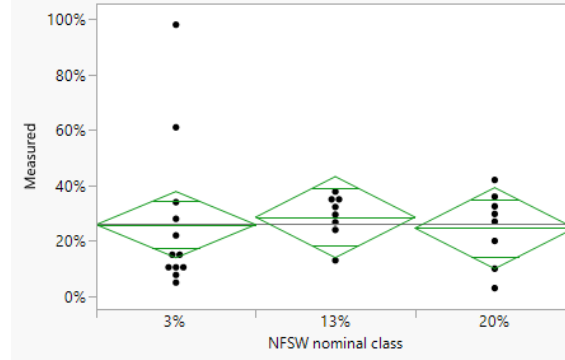
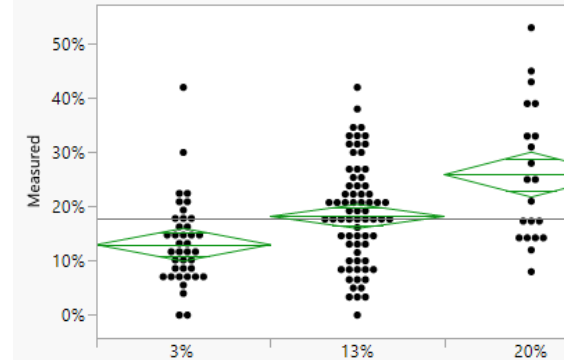


- Required to produce evidence that:
 - Forests / forest species are present in the locality
 - Map out CEA; baseline/pre-existing forest; non-project area
 - Management changes occurring to support regeneration
 - Regeneration is meeting 5-yearly targets:
 - Minimum canopy cover or change thresholds
 - Minimum stocking of appropriate species
 - **Good practice mapping with minimum 85% accuracy:**
 - Satellite with 1 – 10 m resolution; aerial photographs; LiDAR
 - Local calibration / ground truthing
- Get audited and reviewed
 - ...multiple times

What does CER do?

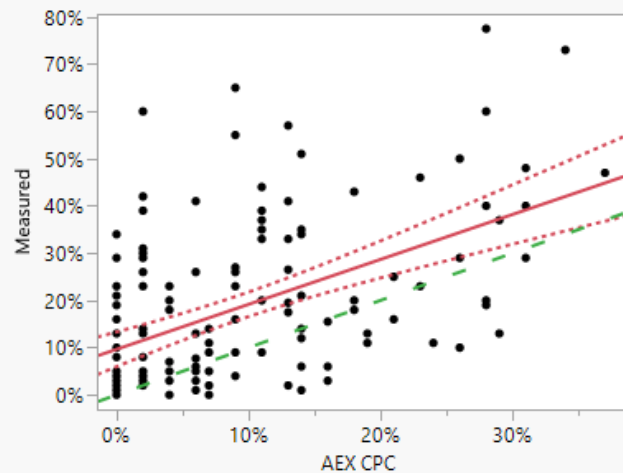


Is NFSW a “useful” model? (b)

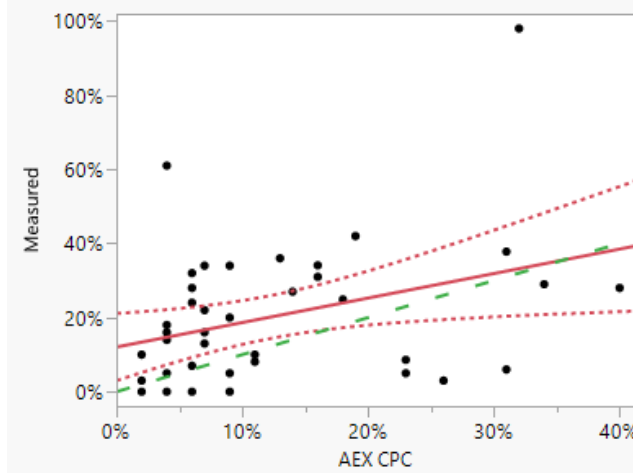
Queensland	New South Wales	Western Australia																																																																								
 <table><tr><th>Source</th><th>DF</th><th>Sum of Squares</th><th>Mean Square</th><th>F Ratio</th><th>Prob > F</th></tr><tr><td>NFSW nominal class</td><td>2</td><td>0.4795977</td><td>0.239799</td><td>8.6592</td><td>0.0003*</td></tr><tr><td>Error</td><td>131</td><td>3.6277727</td><td>0.027693</td><td></td><td></td></tr><tr><td>C. Total</td><td>133</td><td>4.1073704</td><td></td><td></td><td></td></tr></table>	Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F	NFSW nominal class	2	0.4795977	0.239799	8.6592	0.0003*	Error	131	3.6277727	0.027693			C. Total	133	4.1073704				 <table><tr><th>Source</th><th>DF</th><th>Sum of Squares</th><th>Mean Square</th><th>F Ratio</th><th>Prob > F</th></tr><tr><td>NFSW nominal class</td><td>2</td><td>0.0066684</td><td>0.003334</td><td>0.0827</td><td>0.9209</td></tr><tr><td>Error</td><td>25</td><td>1.0080515</td><td>0.040322</td><td></td><td></td></tr><tr><td>C. Total</td><td>27</td><td>1.0147199</td><td></td><td></td><td></td></tr></table>	Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F	NFSW nominal class	2	0.0066684	0.003334	0.0827	0.9209	Error	25	1.0080515	0.040322			C. Total	27	1.0147199				 <table><tr><th>Source</th><th>DF</th><th>Sum of Squares</th><th>Mean Square</th><th>F Ratio</th><th>Prob > F</th></tr><tr><td>NFSW nominal class</td><td>2</td><td>0.2297976</td><td>0.114899</td><td>12.5056</td><td><.0001*</td></tr><tr><td>Error</td><td>130</td><td>1.1944099</td><td>0.009188</td><td></td><td></td></tr><tr><td>C. Total</td><td>132</td><td>1.4242075</td><td></td><td></td><td></td></tr></table>	Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F	NFSW nominal class	2	0.2297976	0.114899	12.5056	<.0001*	Error	130	1.1944099	0.009188			C. Total	132	1.4242075			
Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F																																																																					
NFSW nominal class	2	0.4795977	0.239799	8.6592	0.0003*																																																																					
Error	131	3.6277727	0.027693																																																																							
C. Total	133	4.1073704																																																																								
Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F																																																																					
NFSW nominal class	2	0.0066684	0.003334	0.0827	0.9209																																																																					
Error	25	1.0080515	0.040322																																																																							
C. Total	27	1.0147199																																																																								
Source	DF	Sum of Squares	Mean Square	F Ratio	Prob > F																																																																					
NFSW nominal class	2	0.2297976	0.114899	12.5056	<.0001*																																																																					
Error	130	1.1944099	0.009188																																																																							
C. Total	132	1.4242075																																																																								
Classification accuracy: Non-woody: 42% Sparse-woody: 50% Woody/Forest: 59%	Classification accuracy: Non-woody: 8% Sparse-woody: 13% Woody/Forest: 75%	Classification accuracy: Non-woody: 10% Sparse-woody: 53% Woody/Forest: 57%																																																																								
Canopy Cover (class mean): Non-woody: 15% Sparse-woody: 23% Woody/Forest: 31%	Canopy Cover (class mean): Non-woody: 26% Sparse-woody: 27% Woody/Forest: 25%	Canopy Cover (class mean): Non-woody: 13% Sparse-woody: 18% Woody/Forest: 26%																																																																								

Is WCF / CPC a useful model?

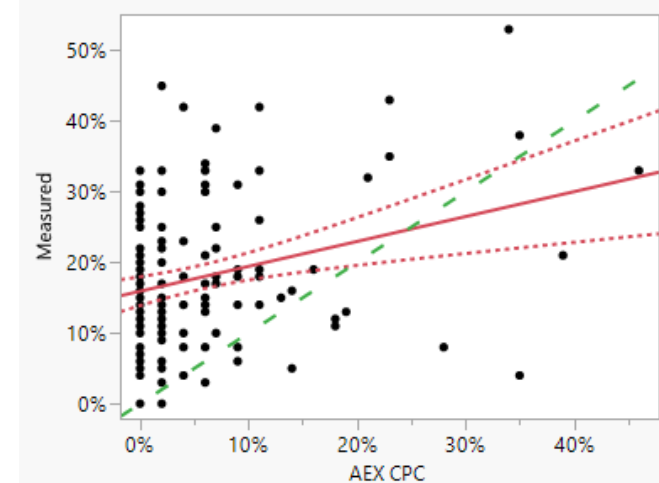
Queensland



New South Wales



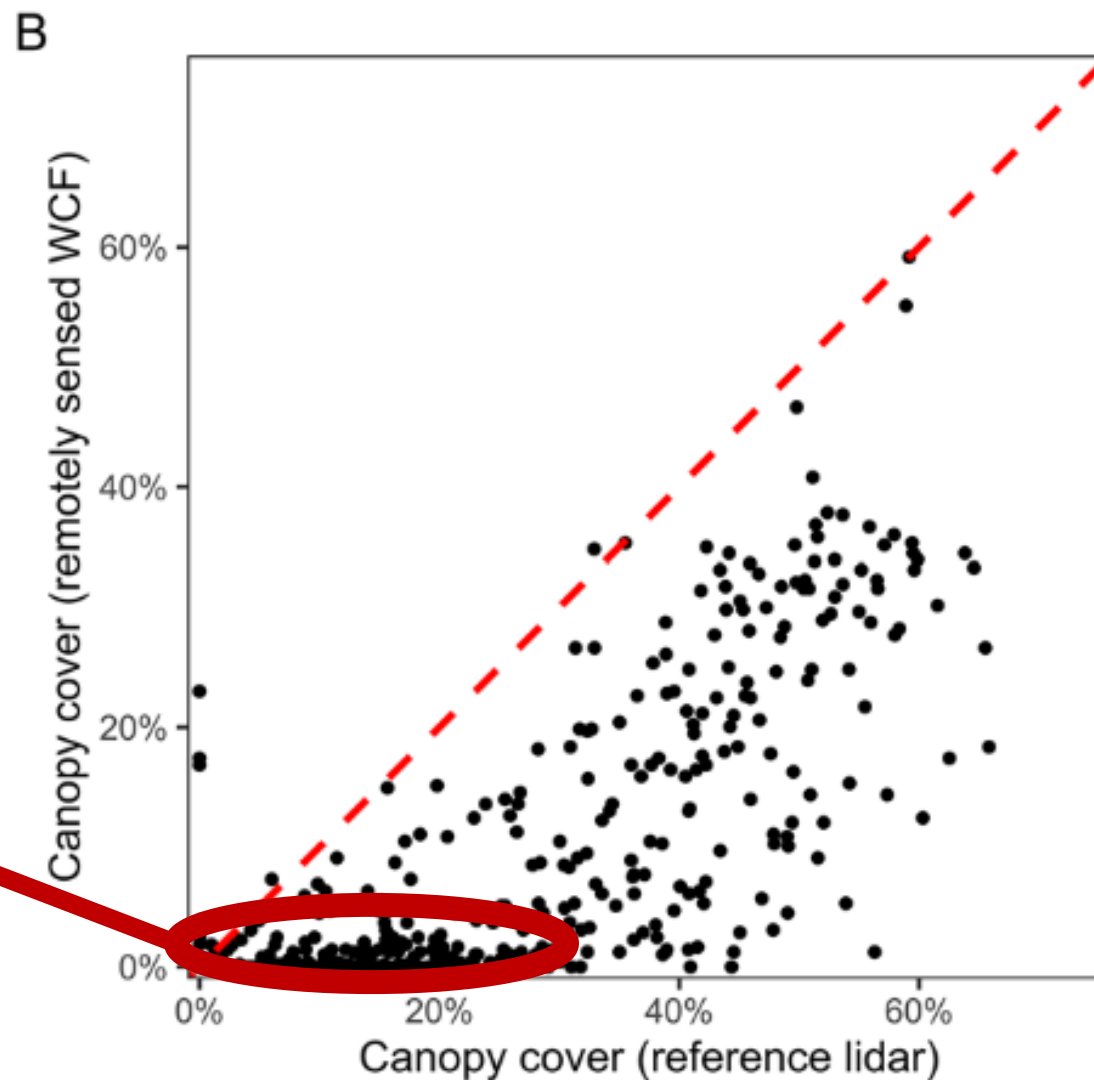
Western Australia



There are significant correlations, but not 1:1 and with RSME more than 20%

Useful as auxiliary variable in multi-phase inventory but NOT to predict canopy cover without correction

WCF without correcting to Canopy?



Moore et al (2025)

Substantial
underestimate / bias in
the lower cover areas



- Strong evidence over-grazing disturbs succession; degrades forest;
- Strong evidence controlling grazing has high probability of returning land to more similar pre-grazing condition;
- National-scale, satellite-based and low precision models unlikely to identify regeneration or degradation;
- High quality data (good resolution satellite; LiDAR; ground-based) and good sampling/inventory essential to monitor change in sparse-woody
- Errors must be adequately dealt with

Sequestering carbon in regeneration: An elaborate hoax or landholders making a difference?

Associate Professor Cris Brack (Honorary)
Cris.Brack@anu.edu.au

Human Induced Regeneration: to generate forests and greater carbon density

