# Utilising airborne scanning laser (LiDAR) to improve the assessment of Australian native forest structure

# Alex C. Lee

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 $River\ Red\ Gum\ (E.\ \textit{camaldulensis})\ located\ on\ an\ island\ in\ the\ Murray\ river,\ NE\ Victoria.$ 

## **TABLE OF CONTENTS**

List of Fig	ures		x
List of Tab	oles		xvi
List of Equ	uations.		xviii
List of Abl	breviati	ons	xix
Certificate	of Aut	horship	xxi
Preface	•••••		<i>xxii</i>
Acknowled	dgemen	ts	xxiii
Abstract	•••••		xxiv
Chapter	1. Inti	roduction	1
1.1	Intr	oduction	1
	1.1.1	The requirement for forest information	1
	1.1.2	Forest assessment in Australia	4
		State of the Forest Reporting	4
		Continental Forest Monitoring	7
	1.1.3	Dealing with scale in remote sensing of forests	10
	1.1.4	Summary	13
1.2	Thes	sis Research Question	15
	1.2.1	Primary research question	15
		Rationale and Research Objectives	15
		Objective 1: Measuring forest cover using LiDAR	16
		Objective 2: Measuring forest height using LiDAR	17
	1.2.2	Research delimitations	18
1.3	Outl	line of Thesis	20
Chapter	2. Res	search issues	21
2.1	Intr	oduction	21

2.2	Con	tinental Forest Measurement in Australia	22
	2.2.1	Description of Australian forests	22
	2.2.2	Defining and measuring Australia's forests	24
		National Vegetation Information System	25
		Current National Forest Inventory reporting	26
		Limitations with NFI forest height and cover reporting	30
	2.2.3	Utilising data within integrated sampling schemes	33
		Sampling strategies overview	33
		Random, systematic, and stratified random sampling	36
		Model based sampling	37
		Using field data for remote sensing calibration	38
2.3	Rem	note Sensing of Forests	40
	2.3.1	Overview	40
	2.3.2	LiDAR remote sensing of forests	46
		Overview	46
		Large footprint LiDAR	50
		LiDAR for Australian forests	51
		LiDAR calibration studies	52
	2.3.3	Tree crown delineation using high resolution remote sensing	54
		Local minima location	55
		Minima contouring	56
		Template matching and object-oriented analyses	57
		Delineation limitations	58
2.4	Scal	e in Remote Sensing	60
	2.4.1	Overview	60
		Modifiable Areal Unit Problem	62
		Hierarchy theory and Landscape Ecology	63

2.5	Sum	mary	65
Chapter 3	. Dat	a Analysis Methodology	67
3.1	Intr	oduction	67
3.2	Rese	earch Design Overview	69
	3.2.1	Multi-scale strategy overview	69
		Developing a hierarchical multi-scale modelling framework	69
		Applying the multi-scale hierarchical framework	71
3.3	Data	Collection	75
	3.3.1	Introduction	75
	3.3.2	Queensland study site	75
		Overview of Queensland multi-stage sampling	78
		Stage III: LiDAR data capture	79
		Stage IV: Field sampling	81
		Stage V: Georeferencing of photography to LiDAR	83
	3.3.3	NE Victorian study site	85
		Plot location site descriptions	88
	3.3.4	Data descriptions for both sites	91
		Ancillary data	91
		Summary of Queensland and NE Victorian field data	91
		API classification comparisons with NFI	91
3.4	Phas	se 1 – LiDAR Plot / Stand Scale Forest Structure Assessment	94
	3.4.1	Introduction	94
	3.4.2	LiDAR pre-processing	94
		Stage VI: Bare ground surfaces	94
		Site Characteristics and terrain complications	96
	3.4.3	Estimating Tree Height for Plot / Stand	98
		Stage VIII: Maximum and predominant height from LiDAR	98

		Plot scale LiDAR apparent vertical profiles	99
		Creating apparent vertical profiles using field data	102
		Growth stage assessment using apparent vertical profiles	103
	3.4.4	Estimating plot scale canopy cover	104
		Stage VIII: Foliage and Crown cover	104
3.5	Phas	se II –Tree and Component Scale Structure Modelling	106
	3.5.1	Tree scale modelling strategy	106
		HSCOI Stage IX: Calculation of stem diameter	107
	3.5.2	Individual crown segmentation and delineation	109
		Conceptual overview	109
		Stage I - Empirical functions for general crown templates	110
		Stage II - Creating individual crown segments	114
		Stage III - Classification of crown segments	116
		Stage IV - Creating crown objects	121
	3.5.3	Tree component scale LiDAR modelling	127
		Branch radius calculation	130
3.6	Phas	se 3 – Multi-Scale Calibration and Validation Case Studies	131
	3.6.1	Introduction	131
	3.6.2	Multi-scale assessment of height and cover	131
		Multi-scale predominant height assessment	131
		Multi-scale canopy cover assessment	136
		LiDAR and API crown cover comparison	137
		Crown Separation method test	138
	3.6.3	Landsat cover investigation and validation	139
	3.6.4	ICESat calibration with airborne LiDAR	142
		Vegetation comparisons	143
		ICESat footprint derivation and attribute extraction	143

3.7	Sum	mary	)
C <b>hapter</b> 4.1		oduction	
4.2	Field	l Plot Representativeness153	;
4.3	Mult	ti-Scale Height Results155	;
	4.3.1	Plot scale height results	;
		Maximum canopy height	ī
		Predominant canopy height	7
	4.3.2	LiDAR height results with different assessment areas	)
		Multi-scale variation of LiDAR height in NE Victoria 159	)
		Multi-scale variation of LiDAR height at Injune 164	!
	4.3.3	Apparent vertical profiles	)
		Simulating apparent vertical profiles	í
		Growth stage and disturbance assessment using profiles 168	}
4.4	Mult	ti-scale Canopy Cover results176	,
	4.4.1	Plot scale cover results	,
		Comparisons between cover using field data	í
		Crown cover comparison between LiDAR and photography 180	)
		LiDAR foliage-branch cover comparisons with field data	)
		SLATS foliage projective cover comparison with LiDAR cover 184	!
		Correlations between cover metrics and data sources	í
		Crown separation method test results	)
	4.4.2	Forest cover assessment at a range of scales	)
4.5	Tree	Scale Stem and Crown Delineation Results 198	;
	4.5.1	Tree stem density and location	,
		Stem mapping at Injune	}
		Tree scale stem diameter derived from height	)

		4.5.2	Tree crown delineation results	201
	4.6	Mult	i-Scale Calibration Results	209
		4.6.1	Landsat pixel scale LiDAR cover calibration	209
		4.6.2	Crown and foliage cover translation function	211
		4.6.3	ICESAT case study calibration using LiDAR height and cover	215
		4.6.4	Stand reconstruction results using tree components	218
	4.7	Sum	mary	225
Cha	pter 5	. Disc	cussion	227
	5.1	Intro	oduction	227
	5.2	Impr	oving the Assessment of Forest Structure using LiDAR	228
		5.2.1	Improving forest structure measurement using LiDAR	228
		5.2.2	Improving forest height assessment using LiDAR	233
			Maximum canopy height	233
			Predominant canopy height	234
			Apparent Vertical Profiles	236
			Maximum stand height at a range of assessment scales	239
			Predominant Height at a range of assessment scales	240
			National Forest Inventory Reporting	241
		5.2.3	Improving forest cover assessment using LiDAR	241
			Field data comparisons	242
			LiDAR cover comparisons with field and CASI	243
			LiDAR to API comparisons	245
			LiDAR to Landsat TM comparisons	247
			Landscape sampling	248
			Multi-scale sampling and reporting of cover	249
		5.2.4	Assessment of the Height Scaled Crown Openness Index (HSCOI)	251
			HSCOI derived stem density	251

		HSCOI derived stem height and diameter2	253
		HSCOI crown delineation2	253
	5.2.5	Calibration case study examples using LiDAR	256
		Landsat TM calibration using LiDAR structural output2	256
		Foliage-branch cover-to-crown cover translation function	257
		ICESat calibration for continental monitoring2	258
		Stand reconstruction modelling for SAR simulation2	259
5.3	Prac	etical Implications, Limitations and Recommendations 2	262
	5.3.1	Practical implications	262
	5.3.2	Limitations and recommendations for future research	264
Chapter	6. Cor	nclusion2	267
6.1	Thes	sis Conclusion2	267
Referenc	es		269
Appendi	x A		289
Ancil	lary Res	sults	289
Appendi	x B		320
Plot	data illus	strations for NE Victoria3	320
Appendi	x C		326
List	of Publis	hed Papers3	326
State	ment of	Author Contribution to Published Papers 3	327
Appendi	x D		332
Hemi	ispherica	al photo calibration methodology overview3	332

# LIST OF FIGURES

Figure 1: CFMF multi-tier design illustrating the potential role of airborne LiDAR as a Tier 2 dataset (after Wood, <i>et al.</i> , 2006)
Figure 2: Distribution of NFI forest types in Australia. Source – State of the Forests Report, 2003 (NFI, 2003)
Figure 3: Illustration of different cover measurement sensors and range of scale and spatial resolution, both field and remotely sensed (McDonald, <i>et al</i> 1998; McCloy, 2006)
Figure 4: Graphical representation of the elements of a small footprint airborne scanning laser (LiDAR) system. GPS refers to Global Positioning System, INS is Inertial Navigation System. Note that the graphic and elements are illustrative only and not to scale
Figure 5: Methods chapter layout showing major components of LiDAR assessment of forest structure
Figure 6: Thesis conceptual overview outlining linkages between multi-scale hierarchal modelling and applications for modelling forests with LiDAR72
Figure 7: Flowchart of multi-scale calibration modelling and application strategy74
Figure 8: Location of the 37 x 60 km Injune study area, within south-central  Queensland
Figure 9: Sketch of the Maranoa River by the explorer Thomas Mitchell in 1846 (upper)
(Mitchell, 1846); and a photo of the Maranoa from a nearby location in
2004 (lower)
Figure 10: Landsat ETM image (2003) of study area and systematic photo and LiDAR
plot layout, with field plot locations circled. State Forests are the light
green hashed areas. Study area boundary was 37 km x 60 km, and
sampling units are 4 km apart80
Figure 11: True colour 1:4000 stereo aerial photograph of PSU 138 overlain with the
500 x 150 m primary sampling unit and 30 secondary sampling units (50
x 50 m). Polygon vectors mapped through API delineate different forest
communities (based on dominant species composition and cover)84
Figure 12: Location of the secondary study site within the Broken and Ovens
catchments in NE Victoria
Figure 13: Layout of LiDAR data collection with ancillary NVIS vegetation data aggregated to broad species classes (NLWRA, 2001) within the NE
Victorian study area87
Figure 14: Field plot site information summary (LiDAR derived) with max tree height
(LiDAR), dominant species groups (field), and broad landform zone
(derived) (Wood <i>et al.</i> , 2006)
Figure 15: Original DEM TIN created with all ground returns, for PSU 14295
Figure 16: Final spatially refined DEM TIN for PSU 14295
Figure 17: DEM difference surface - original DEM subtracted from final DEM for PSU 14295
Figure 18: LiDAR representations – raw points profile across Qld plot width (p81-11),
vertical profile of summed returns per 1m height interval, and cumulative
height percentage curve summary
Figure 19: Steps to generate foliage branch cover, and CHM crown cover from LiDAR
105

• • • • • • • • • • • • • • • • • • • •	Height-to-D <sub>130</sub> translation function using 80% of field measured stems
	and (b) validation using remaining 20% of field stems
-	· · · · · · · · · · · · · · · · · · ·
	and applied to LiDAR derived stems
	vchart of crown delineation methodology
	bration using 80% of field data trees (Upper) and validation using 20%
	of field data trees (Lower) for estimating crown area from height, for (a)
	Eucalypt and Angophora trees, and (b) Callitris and Acacia trees (live
	rees 5cm+ D <sub>130</sub> )
-	wn segmentation using surface generated from a 1m circular (a) and
	5x5m rectangular (b) moving window
_	othing of segmentations (a) and segment clipping with HSCOI derived
	crown boundary delineation (b)
_	tration of different apparent vertical profiles for the two main structural
	ypes
	mples of the two broad genus groups based mature tree structural crown
	forms
Figure 28: Crov	wn structural classification into two broad genus categories at (a)
	segment <sub>1m</sub> and (b) segment <sub>5m</sub> scales. Green is <i>Callitris-Acacia</i> and
1	ight brown is <i>Eucalypt-Angophora</i> . 119
Figure 29: Deri	vation of circular crown shape assumption, using all field live stems
4	5cm + D130 (n = 2708) and comparing crown north-south length versus
•	east-west length. 122
Figure 30: Fina	l crown structural classification (a) and delineation after spatial
8	assessment (b)
Figure 31: Vox	el groups associated with branch clusters within an individual E.
	populnea tree from PSU 142, with a planimetric view (upper) and 3D
_	view (lower)
Figure 32: Pred	ominant stand height assessment using LiDAR 10m cells at different
S	spatial extents at CFMF plot 212 (Ovens river)
	tration of CFMF plot p212 with plot, transect and stand scales of
_	assessment. The LiDAR point data slice was approx. 100m deep133
	ti-scale assessment of predominant height for NE Victorian sites, with
	example for plot 212 shown
	mple of predominant height assessment at a range of scales for Injune
	(PSU 142 shown, with 10m cells within 30 larger 0.25ha SSU's)135
	tration of the multiple scales of FBC circular assessment for NE
	Victorian plots. LiDAR (2m+) for plot 212 has red/orange colour as
	nighest returns (~ 40 m), with lowest non-ground dark blue.
	Background image was Landsat ETM
	tor pixel layer example for PSU 142, derived from Landsat SLATS FPC
-	grid (background). Values within cells indicate FPC141
	AR return density (all returns) for PSU 142. Values within cells indicate
	LiDAR FBC
	COI derived stems for PSU 142. Values within cells indicate stem
_	density per FPC pixel (stems per hectare)141
	COI derived crown delineations for PSU 142. Values within cells
_	ndicate LiDAR crown cover percent per pixel area141
	Sat transects (light blue) with the airborne LiDAR transects (yellow) and
_	numbered overlap locations across CFMF pilot region in NE Victoria.
1	142
	142

Figure 42: Examples of the ICESat footprint size and shape attributes derived using airborne LiDAR. a) single footprint with returns that are within the
, 6
footprint (black) and over 2m in height (light blue); and b) two footprints
from different dates showing different shapes due to different laser
sensors used
Figure 43: Layout of the Results chapter showing the major components of the strategy
for using LiDAR to improve forest structure measurement
Figure 44: Comparison of the LiDAR structural (height and cover) range of field plots and all SSU's
Figure 45: Comparison of maximum height between field and LiDAR at Injune for; (a)
field plots; and (b) individual tree height for 100 manually selected,
isolated overstorey trees
Figure 46: Comparisons field and LiDAR for NE Victorian plots for; a) max plot height
for field plot and transect areas; and b) frequency distribution of
maximum LiDAR height
Figure 47: Frequency distribution of LiDAR maximum height at Injune for: (a) field
plots; and (b) all 4500 SSUs
Figure 48: Injune field plot predominant height (10m cells) for: (a) from field data and
LiDAR at different sampling rates, and (b) frequency distribution of
LiDAR heights
Figure 49: NE Victoria field plot predominant height (10m cells) for: (a) from field data
and LiDAR at different sampling rates, and (b) frequency distribution of
LiDAR heights
Figure 50: Frequency distributions for predominant height at Injune for 4500 SSUs
derived from transfer function using 90 <sup>th</sup> percentile LiDAR height159
Figure 51: Representativeness test for field plot (~0.09 ha) and transect area (~1 ha)
versus swath overlap area (~25ha) for LiDAR (a) predominant height,
and (b) maximum height, in NE Victoria
Figure 52: LiDAR predominant height assessment at a range of spatial scales for NE Victorian sites where field plots are located
Figure 53: Comparison of sampling strategies for estimating stand (~20ha) predominant
height for field plots in NE Victoria using (a) single locations (field or
LiDAR); and (b) multiple samples within stand
Figure 54: LiDAR predominant height assessment at a range of spatial scales for 12
PSUs at Injune. Dominant species codes are given in Chapter 3165
Figure 55: Comparisons between apparent vertical profile from LiDAR and field cubic
modelling for matches that are: (a) good (p142-02), and (b) not as good
(p81-11)167
Figure 56: Correlation of percentage of cubes at each 1m height interval, with LiDAR
and field apparent vertical profiles for matches that are; (a) good (p142-
02), and (b) not as good (p81-11).
Figure 57: Field data summary for SSU 124-19, illustrating tree growth stage and genus
distribution by stem diameter (total Basal Area = $4.30$ m <sup>2</sup> ) and stocking ( $n$
= 603)
Figure 58: Apparent vertical profiles from LiDAR (black) and field (grey) cubic
modelling for SSU 124-19
Figure 59: Growth stage example using indicative LiDAR vertical profiles from NE
Victoria. For the LiDAR profiles the x-axis is percentage of canopy
returns, y-axis is height above ground (m)171
Figure 60: LiDAR apparent vertical profile showing the potential difference in fire
intensity between two high country plots on steep slopes. (a) plot 562,
less intensity fire impact, with understorey and lower canony still

present. (b) plot 558, intense fire impact, no understorey or lower canopy left and scorched crowns
Figure 61: Comparison of LiDAR apparent vertical profiles and plot photos for plots in NE Victorian <i>E. radiata</i> forests, illustrating a potential understorey fire
recovery sequence, based on the last recorded fire within; (a) 1 year (p313), (b) 12 years ago (p550), and (c) 64 years ago (p463)173
Figure 62: Assessment of stand structure using apparent vertical profiles at a range of
scales, for the Injune study area
Figure 63: Injune field data comparisons for transect FPC versus transect FBC (left), and transect FBC (2000) versus field hemi-photo FBC (2004) (right).
Figure 64: Comparison of FBC field plot measurements: a) transects in 2000 and 2004; and b) transects and hemispherical photo in 2004
Figure 65: Estimated hemispherical photo view extent, based on calibration results for
p142-13. HSCOI crown delineations are shown for reference. The field
plot boundaries and transects are approximately 50m long
b) with hemispherical-photo FBC
Figure 67: NE Victorian plot tree-map CC correspondence with hemispherical photo
FBC179
Figure 68: Injune API CC (mid-point of class) correspondence with the LiDAR CC
sample within the API polygon
Figure 69: Comparison of individual API polygon CC (mid-point of class) and LiDAR CC (mean, min and max) based on 25m cells within the PSU. Individual
polygon IDs have been removed for clarity181
Figure 70: NE Victorian API CC (mid-point of class) correspondence with the LiDAR
CC field plot and transect area sample within the API polygon 182
Figure 71: Comparisons of LiDAR FBC at 0.5 and 2 m height thresholds, with a) field
transect FBC; and b) FBC derived from hemispherical-photographs.
Figure 72: LiDAR FBC distribution from a) Injune field plots; and b) 4500 SSUs across study area
Figure 73: SLATS Landsat derived FPC distribution for the Injune study area (220,000ha). NFI forest classes are shown
Figure 74: SLATS Landsat derived FPC distribution for the NE Victoria study area.
NFI forest classes are shown
Figure 75: Crown separation transect method for p142-13. Field mapped stems are
solid orange circles (proportional to $D_{130}$ ) and open circles (proportional to mean crown radius). LiDAR point sampling density (grey) and
HSCOI crown delineations (green) are shown. Field plot boundaries and
transects are 50m long
Figure 76: Crown separation transect method for p81-16. Map elements are described
in the Figure 75 caption. Field plot boundaries and transects are 50m
long
Figure 77: Apparent vertical profiles for p142-13 (a) and p81-16 (b), highlight different vertical foliage characteristics, which may contribute to the different
cover results observed. 191
Figure 78: Comparison of FBC from LiDAR and hemispherical photos, in a) field plot area; and b) transect area
Figure 79: Comparison of FBC from LiDAR in transect area and field plot
hemispherical-photo.

Figure 80: Comparison of FBC for different data and pixel sizes, within the LiDAR swath for NE Victorian plot 220
Figure 81: LiDAR FBC at plot locations with increasing assessment area, for Floodplain ecozone
Figure 82: LiDAR FBC at plot locations with increasing assessment area, for Foothills ecozone
Figure 83: LiDAR FBC at plot locations with increasing assessment area, for Subalpine ecozone
Figure 84: LiDAR FBC at plot locations with increasing assessment area, for the NE Victorian Montane ecozone
Figure 85: LiDAR FBC at plot locations with increasing assessment area for Injune.
Figure 86: Crown/clusters and stem locations identified using the HSCOI surface generated for PSU 142. Darker areas in the HSCOI <sub>stand</sub> surface indicate crowns that are taller and contain a greater density of canopy elements. Internal squares are SSU field plot locations numbered (from left to right) as 02, 13, 18 and 20
Figure 87: Correspondence between field-measured D <sub>130</sub> , and height derived D <sub>130</sub> from HSCOI derived stems
Figure 88: Correspondence between plot-level basal area, for stems measured in the field and estimated from LiDAR HSCOI modelling201
Figure 89: Correlation between field estimated crown area and area derived from LiDAR HSCOI crown delineations, aggregated into broad class types.
Figure 90: Comparison of crown diameter for; a) field and HSCOI trees from plots 142-13 and 81-16, and b) NE Victorian plot CC comparison between field tree-map buffer and HSCOI crowns
Figure 91: NE Victorian LiDAR crown delineation examples and associated field plot centre photos. (a) Plot 382 with a good match between plot level crown cover and stem density; and (b) Plot 562 with a poor match for plot level crown cover and stem density
Figure 92: Comparisons of SLATS FPC and LiDAR FBC at 0.5m (left) and 2m (right) thresholds
Figure 93: Comparison of SLATS FPC and LiDAR CC210
Figure 94: Comparisons of LiDAR stem density and FBC at 0.5m (left) and 2m (right) thresholds
Figure 95: Comparisons of LiDAR stem density and Landsat FPC (left), and LiDAR CC (right)210
Figure 96: Comparison between LiDAR CC and SLATS FPC within 12 PSU's using all 25 m cells where $CC \ge FPC$ ( $n = 855$ )
Figure 97: Derivation of translation function between SLATS FPC and LiDAR CC within 12 PSU's, using 80% of 25 m cells where CC ≥ FPC (n = 683).
Figure 98: Validation of translation function between SLATS FPC and LiDAR CC within 12 PSU's, using 20 %f 25 m cells ( <i>n</i> = 172) where CC ≥ FPC.
Figure 99: Comparison between LiDAR CC and FBC within 1161 x 25 m pixel sized areas from 12 PSU's
Figure 100: Comparison between LiDAR CC and FBC for Injune SSU's (50 m), NE Victorian field plots (30 m), and selected ICESat footprints from NE Victoria (50 - 100m). Red arrows indicate current NFI CC-FPC translation thresholds

Figure 101: Airborne LiDAR from three ICESat footprints from ALS tile 26 displayed on a 1 m LiDAR derived DEM216
Figure 102: Perspective view of LiDAR within ICESat footprint areas. See Figure 101 for legend
Figure 103: Sample of mapped tree stems, crowns, and primary (thicker lines) and secondary (thinner lines) branch distributions in PSU 142 derived from the LiDAR data. Inset extent is marked as a dashed box219
Figure 104: Perspective view of a portion of PSU 142 showing mapped tree stems, crowns, and voxels derived from the LiDAR analyses. The view extent of Figure 103 is in the foreground
Figure 105: (a) <i>E. populnea</i> -dominated forest at PSU_142. (b) Graphical output of part of the reconstructed canopy from the DSTO radar simulation (Lucas <i>et al.</i> , 2006c)
Figure 106: Injune correspondence between field data and LiDAR allometric and component estimated biomass – all assessed stems ( $n = 200$ )221
Figure 107: Injune correspondence between field data and LiDAR allometric and component estimated biomass – LiDAR modelling outlier stems removed $(n = 185)$
Figure 108: Injune correspondence between field data and LiDAR allometric and component estimated biomass – LiDAR modelling and highly different field outlier stems removed ( $n = 170$ )
Figure 109: Injune correspondence between LiDAR allometric and LiDAR component estimated biomass – all assessed stems $(n = 200)$ 223
Figure 110: Summary of predominant and max height for each CFMF field plot, by ecozone
Figure 111: Comparison of stand sampling strategies for estimating stand (~20ha) predominant height by ecozone
Figure 112: SLATS Landsat FPC (2000) spatial distribution at the Injune study site.
Figure 113: NE Victorian SLATS FPC (uncalibrated) from two Landsat scenes.  Systematic field plots are yellow, additional calibration plots are pink.  306
Figure 114: Injune cover matrix graphs. All scales are percent cover
Figure 115: NE Victoria cover matrix graphs. All scales are percent cover312 Figure 116: LiDAR derived tree stem basal area (per pixel) versus SLATS FPC, for 1114 Landsat pixels in 12 PSUs at Injune319
Figure 117: Registration of calibration images using Nikon fisheye lens, showing effective view area, zenith rings, and gaps in extent with the merged photos
Figure 118: Calibration images taken at ANU showing area imaged, and the same effective view extent with (a) Nikon lens, and (b) Minolta lens334
Figure 119: Effect of background colour and pixel brightness threshold in hemispherical photo analysis on cover estimates, using Nikon and Minolta lenses. 335
Figure 120: Simulated viewsheds at different zenith angles for Nikon fisheye photos when taken along Qld transects within a field plot336
Figure 121: Assessing potential hemispherical photo view areas for NE Victorian plots against of LiDAR returns (FBC) clipped at a range of circular areas.
Figure 122: Assessing view area of NE Victorian plot hemispherical photos against LiDAR (FBC) % circular area assessment for plot and transect areas respectively

# LIST OF TABLES

Table 1: Area of NFI forest types ('000 ha) across the States and Territories in A	
Source: National Forest Inventory State of the Forests Report, 20	003. 27
Table 2: NFI translation between foliage projective cover (FPC) and crown cov	· · · · · · · · ·
(National Forest Inventory, 1998)	
Table 3: Advantages and disadvantages assumptions of different hierarchal pro	
levels for forest assessment using LiDAR	
Table 4: Main stages in the acquisition, processing and analysis of field and rem	notely
sensed data (Tickle, et al., 2006)	
Table 5: Estimated year of last fire and logging for NE Victorian plots (DSE, 20	·
Table 6: NFI, Victorian SFRI, and Queensland API crown cover class standard	
Table 7: Height class and ranges for NFI, Victorian SFRI, and Queensland API	
standards	93
Table 8: Illustration of a selection of terrain complications on canopy height est	
Table 9: Translation between LiDAR and field estimation methods for predomin	nant
height	99
Table 10: Growth Stage type using SFRI and EMC classifications	103
Table 11: Overview of processing stages for the HSCOI	107
Table 12: Genus probability selection criteria	
Table 13: Spatial criteria for merging a segment with neighbouring segments	123
Table 14: Spatial criteria for splitting a segment	126
Table 15: Functions used to calculate branch start and end radius	
Table 16: Multi-scale predominant height assessment buffer selection areas for Victorian plot locations.	
Table 17: Formulae utilised for footprint dimension and area calculations:	
Table 18: ICESat attributes extracted for comparison with airborne LiDAR	
Table 19: Variables extracted or calculated for footprint size estimation	
Table 20: Assignment of LiDAR returns within ICESat footprint	
Table 21: ICESat case study footprints; description, full waveform, and LiDAR	
apparent vertical profiles	
Table 22: Description of LiDAR attributes utilised for ICESat calibration and va	alidation
Table 23: Representativeness of Injune forest structure sampling, comparing Lil	
from 4500 SSUs across the landscape with LiDAR from field plo	ots (in
parentheses), by NFI class.	154
Table 24: Representativeness of NE Victorian forest structure sampling, compar	
continental NFI data with field plot LiDAR (in parentheses)	
Table 25: Maximum height comparison between field and LiDAR for Injune an Victoria plots	155
Table 26: Comparisons of predominant height: field and LiDAR for Injune and Victoria plots	
Table 27: LiDAR height comparisons at a range of assessment area scales for N	
Victoria	
Table 28: Summary statistics for predominant height at a range of assessment so	
ner NF Victorian ecozone	163

Table 29: Summary of matrix results for the field plot area	
Table 30: Plot descriptions for three plots illustrating growth progression with time since fire	
Table 31: Comparison of relative accuracy of CC estimates between API and LiDAR	
CC, across different LiDAR sampling ranges of the API polygon 180	
Table 32: Comparison of NFI forest cover class distribution across the Injune landscape	
using SLATS FPC for study area, and LiDAR PSU sample (1,125ha) for	•
FBC and CC	
photo data in field plot and transect areas, using percentage of 20km	
systematic samples in each NFI forest class	
Table 34: Matrix of correlations for cover metrics across a range of datasets and	
measurement scales, for the Injune field plots *	
Table 35: Matrix of correlations for cover metrics across a range of datasets and measurement scales, for the NE Victorian field plots	
Table 36: Matrix of P-values for cover metric correlations across a range of datasets and	d
measurement scales, for the NE Victorian field plots	u
Table 37: Crown separation test comparison for p142-13 and p81-16189	
Table 38: FBC comparison between field plot and transect area combinations for NE	
Victoria	
Table 39: Summary of LiDAR FBC at different scales, averaged per NE Victorian ecozone	
Table 40: Tree stem density per plot correspondence for Injune and NE Victoria from	
HSCOI modelling	
Table 41: Description of LiDAR crown delineation validation classification types. 202	
Table 42: Summary of crown delineation validation comparison of 90 trees aggregated	
by type	
validation	
Table 44: Genus probably modelling accuracy assessment by delineation type, using 90	)
randomly selected trees	
Table 45: SLATS FPC comparisons with LiDAR derived cover and stem density	
metrics for Injune	
Table 46: LiDAR CC comparisons with SLATS FPC and LiDAR FBC, for Injune and NE Victoria	
Table 47: LiDAR and ICESat attributes from the riparian strip footprint f (ICE-id = 48)	4
)217	
Table 48: LiDAR and ICESat attributes from the mid-slope footprint (ICE-id = 480)	
217	
Table 49: LiDAR and ICESat attributes from the ridge top footprint (ICE-id = 475)	
Table 50: Comparison of structural attributes between field data (4 plots) and LiDAR	
stand modelling for PSU 142	
Table 51: Summary of biomass estimates for PSU 142 plots from field, LiDAR stem	
allometric, and LiDAR component sources	
Table 52: Summary of biomass functions using field data and LiDAR stem and	
components	
schemes	
Table 54: Description of different types of LiDAR and field error with stand	
reconstruction using component modelling, for SAR simulation261	
Table 55: Species name and API codes for Injune trees	

Table 56: Injune field plot height results from field and LiDAR data for max and predominant height, and at different measurement scales290
Table 57: NE Victorian field plot height results across a range of datasets and
measurement scales. All values are in metres
Injune
LiDAR returns (blue), and plot centre hemispherical photo297
Table 60: Injune field plot percentage cover results across a range of datasets, cover metrics, and measurement scales
Table 61: NE Victorian field plot percentage cover results across a range of datasets,
cover metrics, and measurement scales. All values in percentage cover.
Table 62: Crown Separation ratio calculation test for p142-13
Table 63: Crown Separation ratio calculation test for p81-16
Table 64: Slope and intercept significance values for calibration and validation functions
Table 65: LiDAR return density at different height thresholds for Injune plots317
Table 66: LiDAR return density at different height thresholds for NE Victorian plots.
Table 67: NE Victorian field plot data (left to right) - LiDAR CHM surfaces, LiDAR
apparent vertical profiles, and plot centre hemispherical photos. With
CHM surfaces dark blue is ground, red-brown is tallest canopy. Refer to
LiDAR profile for respective heights (m)
LIST OF EQUATIONS
Equation 1
Equation 2
Equation 3 112

## LIST OF ABBREVIATIONS

AIRSAR	Airborne Synthetic Aperture Radar
AGO	Australian Greenhouse Office
API	Aerial Photographic Interpretation
BA	Basal Area
CASI	Compact Airborne Spectral Imager
CC	Crown Cover
CFMF	Continental Forest Monitoring Framework
CHM	Canopy Height Model
CRCGA	Cooperative Research Centre for Greenhouse Accounting
D <sub>130</sub>	Diameter of tree stem at 130 cm height above ground
DEM	Digital Elevation Model
DNRE	Department of Natural Resources and Environment (Victoria)
DPI	Department of Primary Industries (Queensland)
DSTO	Defence Science and Technology Organisation
DTM	Digital Terrain Model
EMC	Ecological Maturity Classification
FBC	Foliage-Branch Cover
FPC	Foliage Projective Cover
GIS	Geographic Information Systems
GPS	Global Positioning System
HSCOI	Height Scaled Crown Openness Index
ICESat	Ice, Cloud, and land Elevation Satellite
IBRA	Interim Biogeographical Regionalisation for Australia
JERS	Japanese Earth Resources Satellite
LiDAR	Light Detection And Ranging
MAUP	Modifiable Areal Unit Problem

MODIS	_MODerate-resolution Imaging Spectroradiometer
NFI	_National Forest Inventory
NFPS	_National Forest Policy Statement
NLWRA	_National Land and Water Resources Audit
NVIS	_National Vegetation Information System
PSU	_Primary Sampling Unit
QDNRM	_Queensland Department of Natural Resources and Mines
RSE	_Residual Standard Error
RWG2	_Research Working Group 2
δ	_Standard Deviation
SAR	_Synthetic Aperture Radar
SFRI	_Statewide Forest Resource Inventory (Victoria)
SLATS	_Statewide Landcover And Trees Study (Queensland)
SLR	_Single Lens Reflex (camera)
SPOT	_Satellite Pour l'Observation de la Terre
SOFR	_State of the Forests Report
SSU	_Secondary Sampling Unit
TIN	_Triangulated Irregular Network
TM / ETM	_Thematic Mapper / Enhanced Thematic Mapper
UTM	_Universal Transverse Mercator

# **CERTIFICATE OF AUTHORSHIP**

I hereby declare that this submission is my own work and that, to the best of my
knowledge and belief, it contains no material previously published or written by another person
nor material which to a substantial extent was accepted for the award of any other degree or
diploma at the Australian National University or any other educational institution, except where
due acknowledgement is made in the thesis. Any contribution made to the research by
colleagues with whom I have worked at the Australian National University or elsewhere during
my candidature is fully acknowledged.
Alex Lee Date:

### **PREFACE**

Parts of the description of the Injune study area in Chapter 3 of this thesis was published in:

Tickle, P. K., Lee, A., Lucas, R. M., Austin, J. and Witte, C. (2006) Quantifying Australian forest floristics and structure using small footprint LiDAR and large scale aerial photography. *Forest Ecology and Management*, 223, 379-394.

Description of the Height Scaled Crown Openness Index (HSCOI) in Chapter 3, 4, and 5 was published in:

Lee, A. C. and Lucas, R. M. (2007) A LiDAR-derived Canopy Density Model for Tree Stem and Crown Mapping in Australian Forests. *Remote Sensing of Environment*, 111, 493-518.

Description and discussion of the LiDAR modelling for stand reconstruction for SAR simulation was published in:

Lucas, R. M., Lee, A. C. and Williams, M. L. (2006) Enhancing SAR simulations using LiDAR for understanding the relations between forest structure and SAR imagery. *IEEE Transactions on Geoscience and Remote Sensing*, 44, 2736-2754.

In Tickle *et al.*, (2006) and Lucas *et al.*, (2006) all LiDAR related processing and discussion was undertaken by me.

Appendix 3 provides a more detailed description of author contributions to the above papers.

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#### **ABSTRACT**

Enhanced understanding of forest stocks and dynamics can be gained through improved forest measurement, which is required to assist with sustainable forest management decisions, meet Australian and international reporting needs, and improve research efforts to better respond to a changing climate. Integrated sampling schemes that utilise a multi-scale approach, with a range of data sourced from both field and remote sensing, have been identified as a way to generate the required forest information. Given the multi-scale approach proposed by these schemes, it is important to understand how scale potentially affects the interpretation and reporting of forest from a range of data.

To provide improved forest assessment at a range of scales, this research has developed a strategy for facilitating tree and stand level retrieval of structural attributes within an integrated multi-scale analysis framework. The research investigated the use of fine-scale (~1m) airborne Light Detection and Ranging (LiDAR) data (1,125 ha in central Queensland, and 60,000 ha in NE Victoria) to calibrate other remotely sensed data at the two study sites. The strategy refines forest structure mapping through three-dimensional (3D) modelling combined with empirical relationships, allowing improved estimation of maximum and predominant height, as well as foliage and crown cover at multiple scales. Tree stems (including those in the sub-canopy) were located using a height scaled crown openness index (HSCOI), which integrated the 3D density of canopy elements within the vertical profile into a two-dimensional spatial layer. The HSCOI modelling also facilitated the reconstruction of the 3D distribution of foliage and branches (of varying size and orientation) within the forest volume.

Comparisons between forests at the Queensland and NE Victorian study sites indicated that accurate and consistent retrieval of cover and height metrics could be achieved at multiple scales, with the algorithms applicable for semi-automated use in other forests with similar structure. This information has facilitated interpretation and evaluation of Landsat imagery and ICESat satellite laser data for forest height and canopy cover retrieval. The development of a

forest cover translation matrix allows a range of data and metrics to be compared at the plot scale, and has initiated the development of continuous transfer functions between the metrics and datasets. These data have been used subsequently to support interpretation of SAR data, by providing valuable input to 2D and 3D radar simulation models. Scale effects have been identified as being significant enough to influence national forest class reporting in more heterogeneous forests, thus allowing the most appropriate use and integration of remote sensed data at a range of scales. An empirically based forest minimum mapping area of 1 ha for reporting is suggested. The research has concluded that LiDAR can provide calibration information just as detailed and possibly more accurately than field measurements for many required forest attributes. Therefore the use of LiDAR data offers a unique opportunity to bridge the gap between accurate field plot structural information and stand to landscape scale sampling, to provide enhanced forest assessment in Australia.