



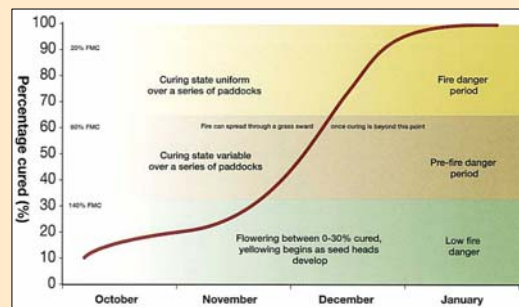
 <div style="float: right; text-align: right;">   </div>	
PROGRAM A	
	<h2 style="margin: 0;">LINKING FIELD OBSERVATIONS WITH REMOTE SENSING TO DETERMINE GRASSLAND FIRE HAZARD</h2> <hr style="width: 50%; margin: 20px auto;"/> <p>Stuart Anderson Ensis Forest Biosecurity and Protection, Scion, Christchurch, New Zealand</p> <p>Elizabeth Botha Ensis Forest Biosecurity and Protection, CSIRO, Canberra</p>
<div style="display: flex; justify-content: space-between; align-items: center;">  <div> <small>© BUSHFIRE CRC LTD 2007</small> <small>THE JOINT FORCES OF CSIRO & SCION</small>  </div> </div>	

	
Program A - Linking field observations with remote sensing to determine grassland fire hazard <small>© bushfire CRC LTD 2007</small>	
	<h2 style="margin: 0;">Outline</h2> <ol style="list-style-type: none"> 1. Grassland curing 2. Project overview 3. Research methods 4. Field data collection 5. Remote sensing



Grassland curing

- Degree of curing - the proportion of dead material in a grassland fuel complex (%)

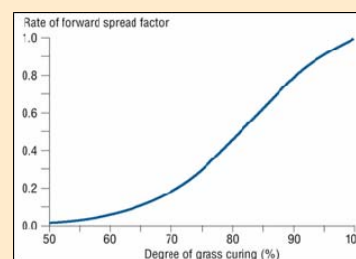
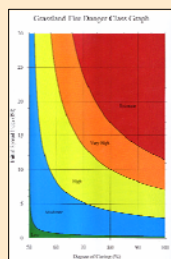


(Source: Garvey and Millie 2001)



Why is curing important?

- Input into fire behaviour models and fire danger rating systems in Australia and New Zealand
- Grassfires significant in both countries
- Prescribed fire also very important
- Often poorly assessed

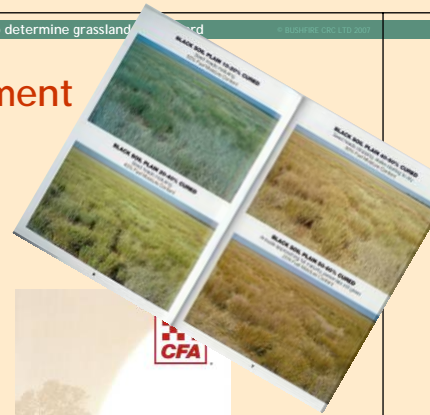
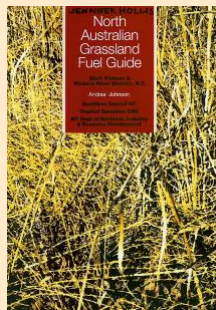
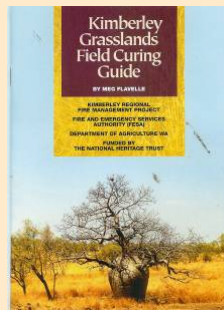




Current curing assessment

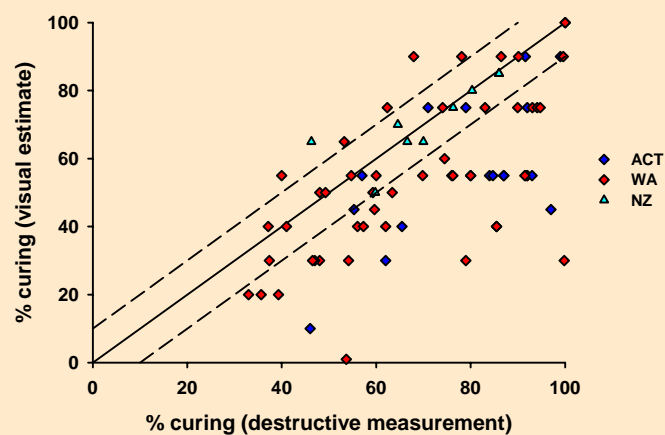
Visual

Poor correlation between
visuals and actual curing



Curing assessment: Visual

Visual vs destructive - 2005-2007





Current curing assessment

Visual

- Poor correlation between visuals and actual curing

Destructive

- Time-consuming
- Difficult

Remote sensing

- Cloudiness
- Pixel size
- Non-uniform fuels
- Discontinuous cover/patchiness
- Outdated technology



Aim:

This project will develop better methods to assess the current and predicted levels of curing in grasslands as an input into fire danger rating systems and fire behaviour models





Research approaches

1. Remote sensing
2. Curing modelling
3. Field sampling

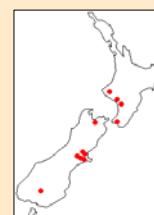
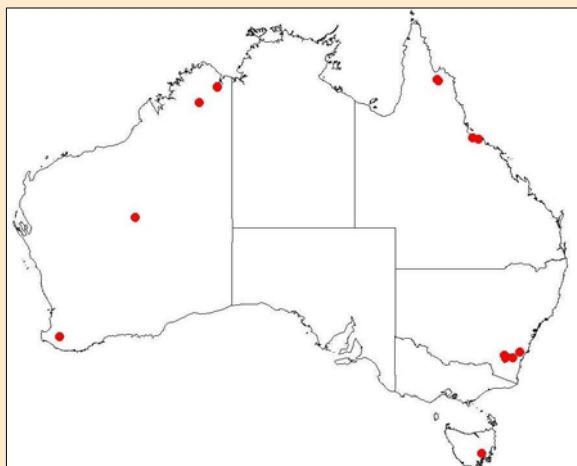


Field data collection

- Field data is ESSENTIAL
- Collected across grasslands of Australia and New Zealand
- A major task, but very important
- Destructive sampling most reliable - cost, effort, time
- Alternative approach - modified Levy Rod



Field Sites



Data collection: Levy rod method

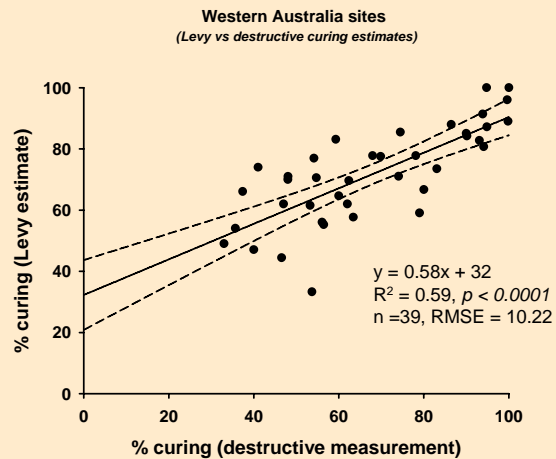


- Total number of live and dead grass touches against a thin steel rod
- Record at intervals along fixed transects
- Less subjectivity

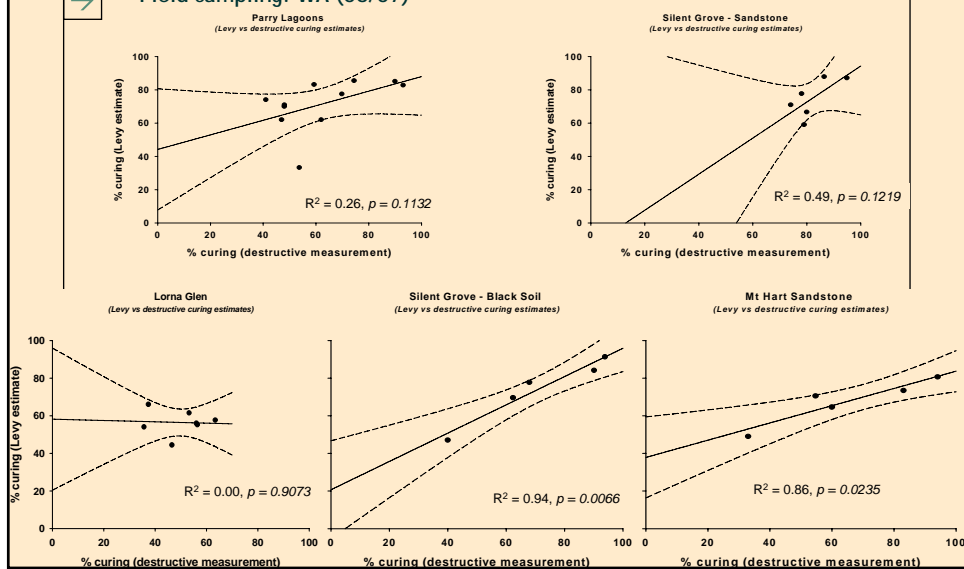




Field sampling: WA (06/07)

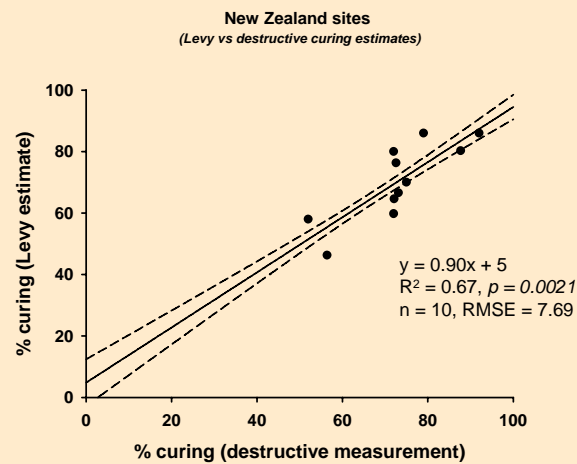


Field sampling: WA (06/07)





Field sampling: NZ (06/07)



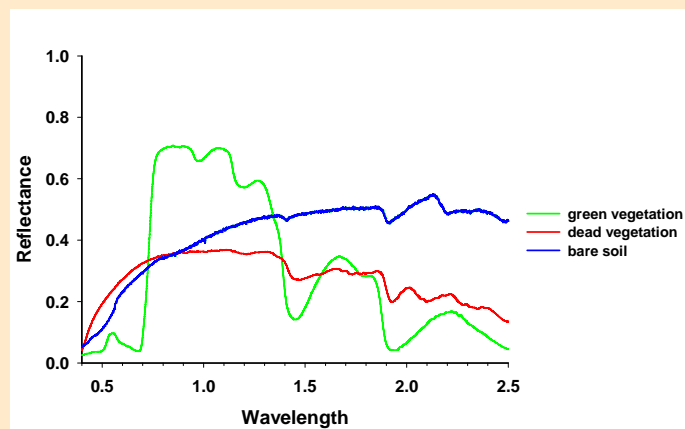
Field Data Collection

1. Levy Rod Assessments
2. Fuel Moisture Content
3. Destructive Sampling (selected sites)

Support from end-user agencies is appreciated



Remote Sensing: Background



Satellite data



MODIS MOD09 image product

- a) 7 spectral bands
- b) 500x500m spatial-resolution
- c) 8-day temporal resolution



Image source: http://terra.nasa.gov/Brochure/Sect_4-7.html

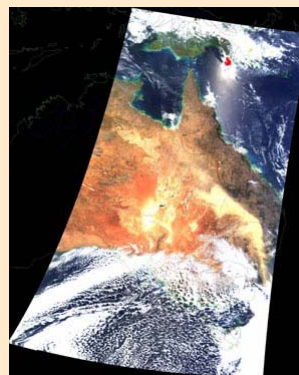
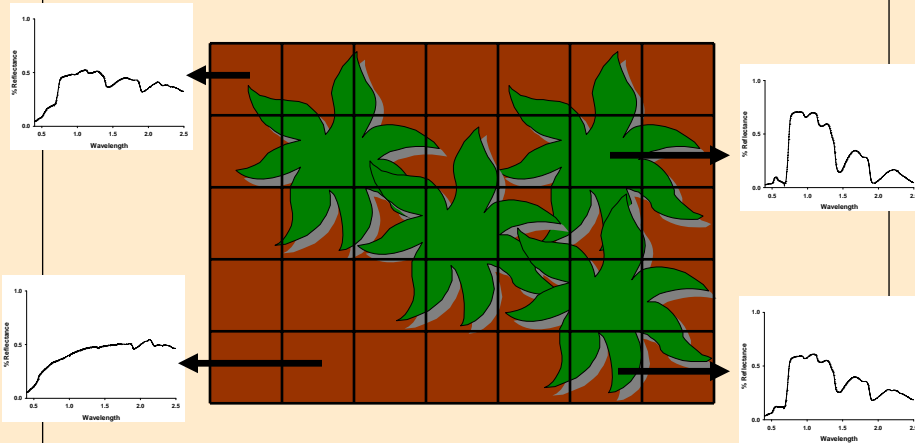


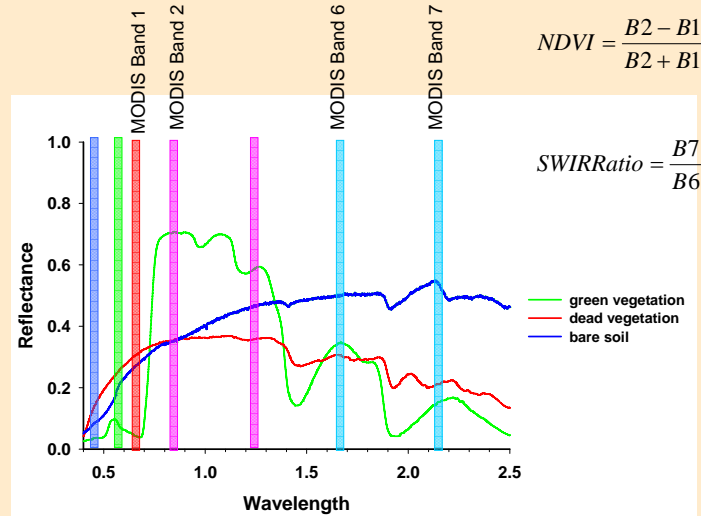
Image source: <http://www.ga.gov.au>



Remote Sensing: What the satellite sees

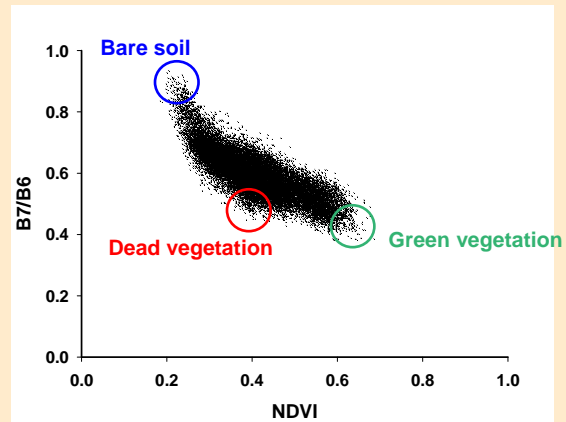


Remote Sensing



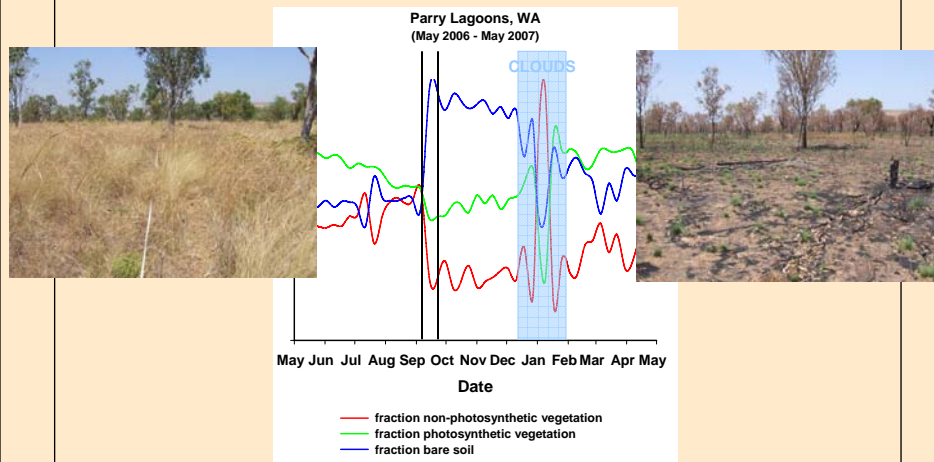


Spectral unmixing

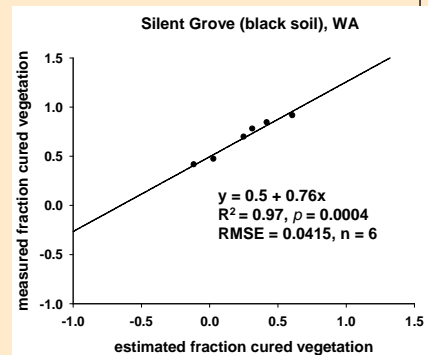
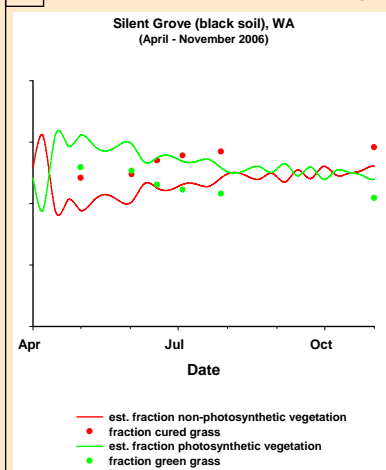




WA Spectral Unmixing

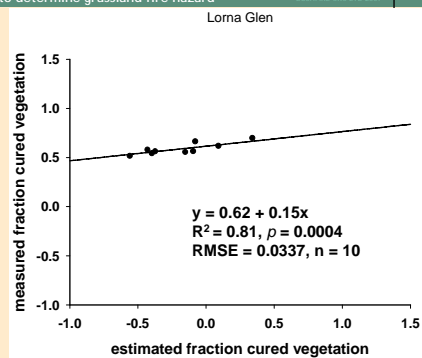
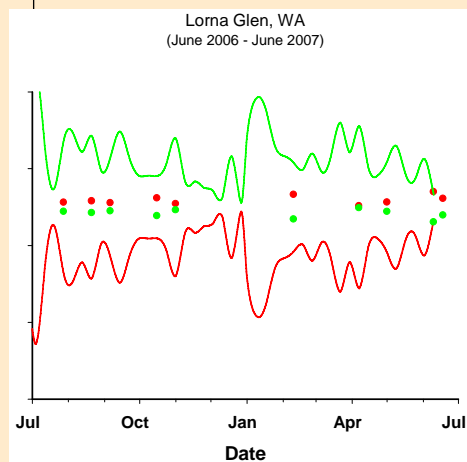


Results: WA (Kimberley)

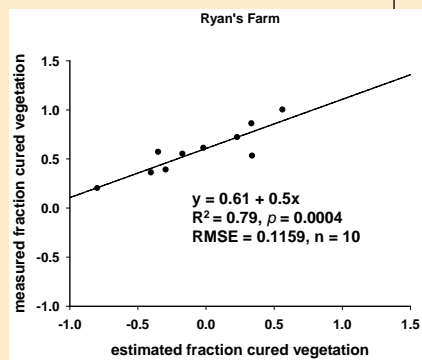
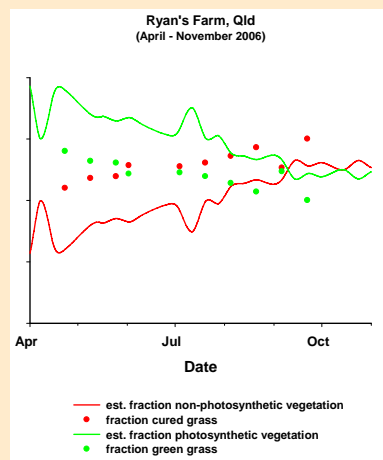




Results: WA (Goldfields)

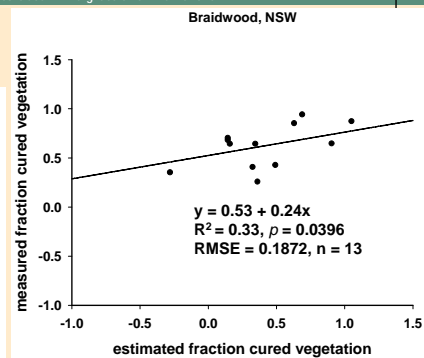
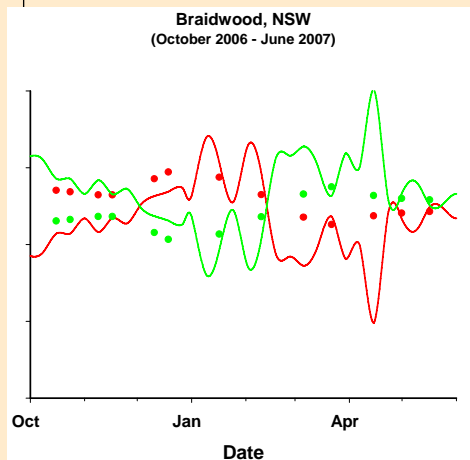


Results: Qld

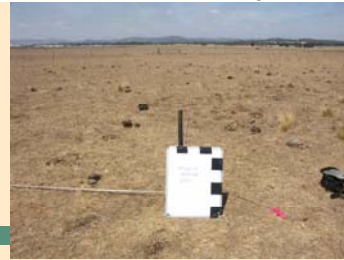
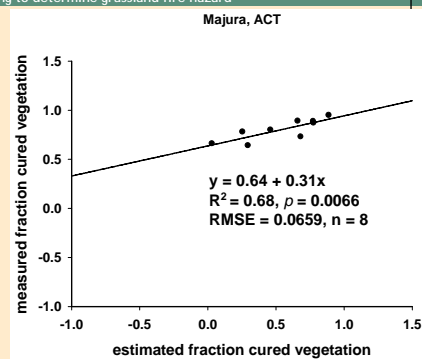
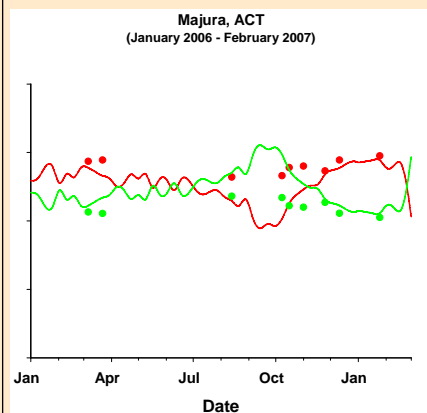




Results: ACT/NSW



Results: ACT/NSW



	<div data-bbox="1112 262 1274 325">bushfire CRC</div>
	<div data-bbox="462 331 1055 357">Program A : Linking field observations with remote sensing to determine grassland fire hazard</div> <div data-bbox="1112 331 1226 357">bushfire CRC 11b 2007</div>
<div data-bbox="389 378 430 430">→</div>	<div data-bbox="462 388 795 430">Preliminary results</div> <ol style="list-style-type: none"> 1. Spectral patterns of dead/live vegetation correlate with ground observations 2. No 1:1 relationship between unmixing results and ground data - model will have to be calibrated separately for each region 3. Best results in tropical savanna area where model was developed 4. Each region may require different end-members to optimize the spectral unmixing algorithm. 5. We need more ground observations to validate this method - especially of the grasslands in the southern part of Australia.

	<div data-bbox="1112 1134 1274 1197">bushfire CRC</div>
	<div data-bbox="462 1203 1055 1228">Program A : Linking field observations with remote sensing to determine grassland fire hazard</div> <div data-bbox="1112 1203 1226 1228">bushfire CRC 11b 2007</div>
<div data-bbox="389 1249 430 1302">→</div>	<div data-bbox="462 1260 803 1302">Acknowledgements</div> <p>Wendy Anderson (UNSW@ADFA) Jennifer Hollis (DEC, WA) Ruth Gibbs (Ensis/CSIRO) Kelsy Gibos (formerly Ensis/Scion) Juan Pablo Guerschman (CSIRO Land and Water) Leigh Douglas (formerly Ensis/CSIRO) Peter Ellis (Ensis/CSIRO) Francis Hines and Adam Fountain (formerly UNSW@ADFA)</p> <p>End-user agencies:</p> <p><u>Australia</u>: Dept of Environment & Conservation (WA), Qld Fire & Rescue Service, ACT Government, private landowners</p> <p><u>New Zealand</u>: Dept of Conservation, National Rural Fire Authority, NZ Forest Owners Assn, Local Government NZ, NZ Defence Force, private landowners</p>