

Regional fluxes of methane in the Surat Basin, Queensland

Stakeholder Roundtable Group meeting

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4 November 2019



Australian Government
Department of Industry,
Innovation and Science



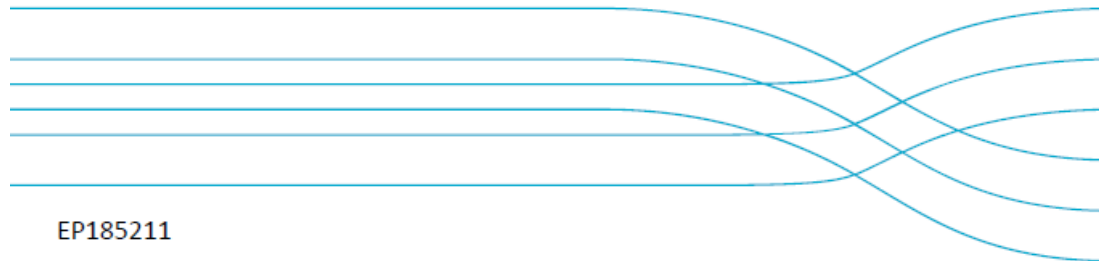
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Characterisation of Regional Fluxes of Methane in the Surat Basin, Queensland

Final report

Task 3: Broad scale application of methane detection, and
Task 4: Methane emissions enhanced modelling



EP185211

Report for the Gas Industry Social and Environmental Research Alliance (GISERA)

October 2018

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Aim of the project

- Demonstrate the utility of an atmospheric “top-down” or inverse modelling approach to infer regional scale ($\sim 100 - 1000$ km) methane emissions across the Surat Basin
- Monitoring from 2 stations: Ironbark and Burncluith (concurrent measurements during July 2015 – December 2016)
- Compare with “bottom-up” inventory emissions

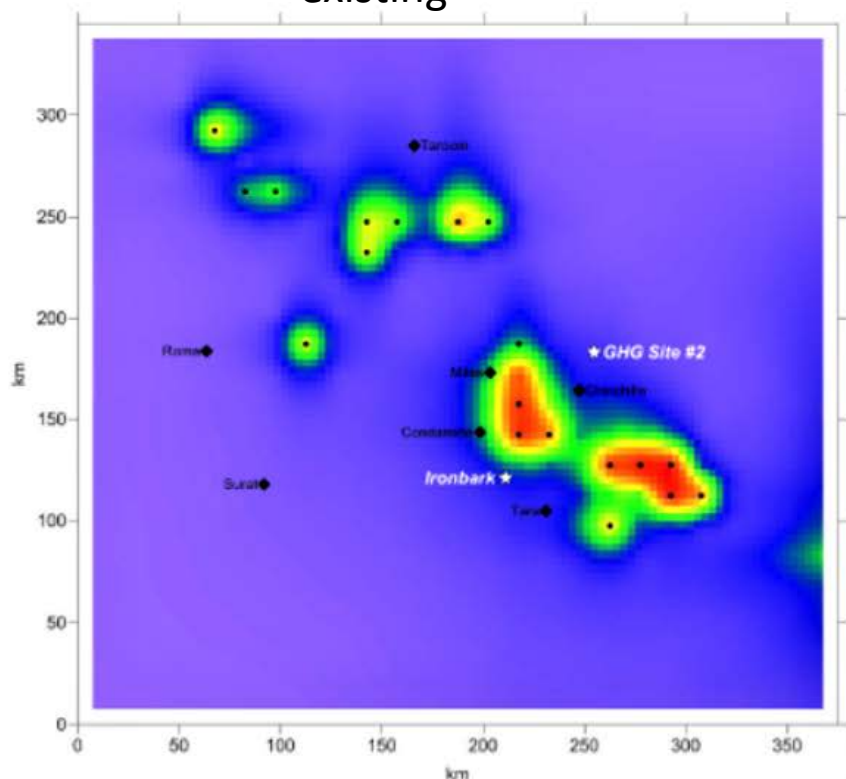
Surat Basin



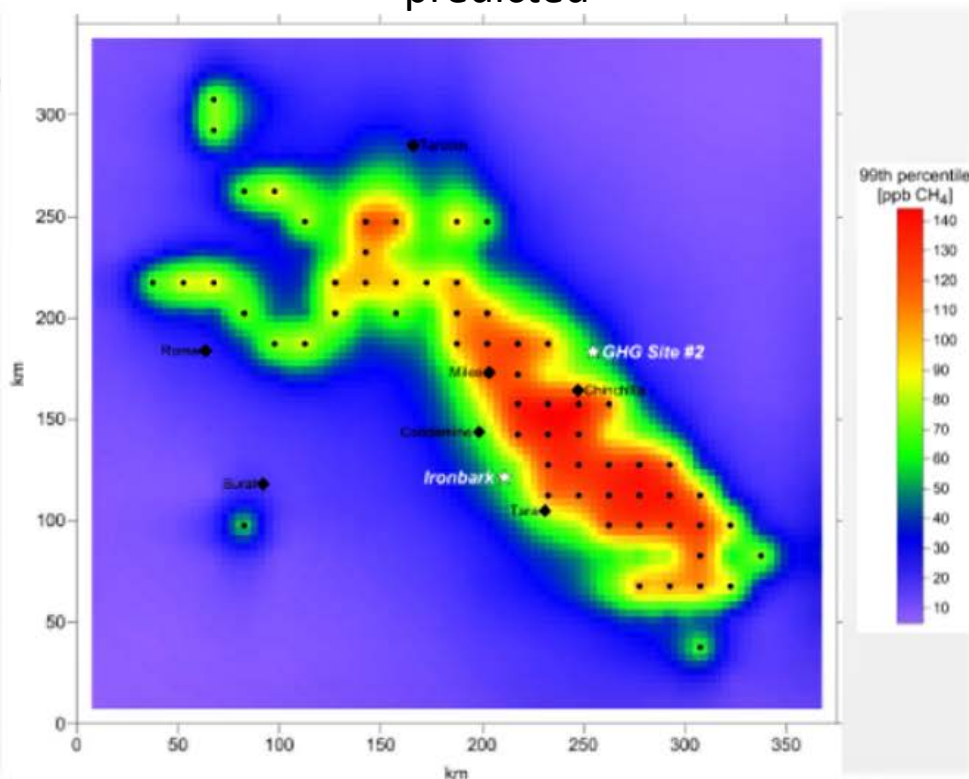
Simulated CH₄ concentrations from CSG wells 2015 – 2018 to optimise monitoring design

Modelled methane concentration signals (TAPM) from CSG operations

existing



predicted



Ironbark (IBA)

CH₄ and CO₂ concentration, meteorology, eddy-covariance fluxes



Burncluith (BCA)

CH₄, CO₂ and CO concentration, meteorology

CH₄ precision of both stations ~0.2%



Technical and environmental challenges



Local engagement



Concentrations at Ironbark

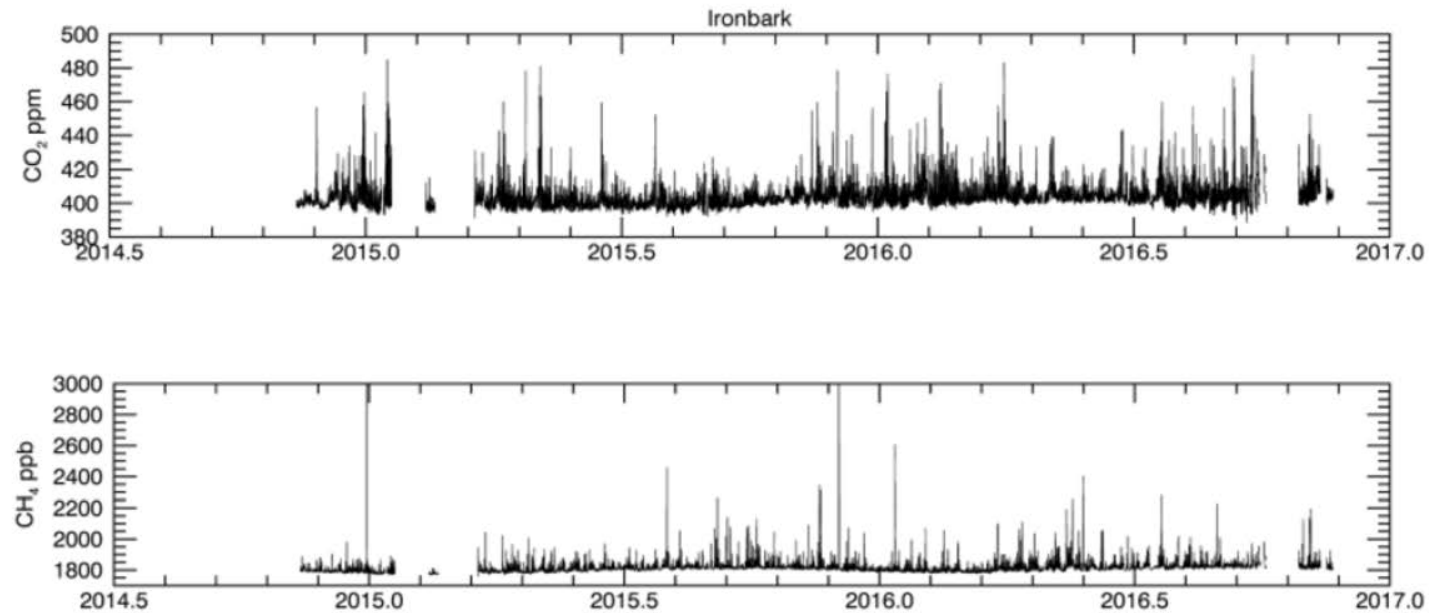


Figure 3. Measured concentration time series (hour means) of CO₂ (parts per million, ppm) and CH₄ (parts per billion, ppb) at Ironbark.

Concentrations at Burncluith

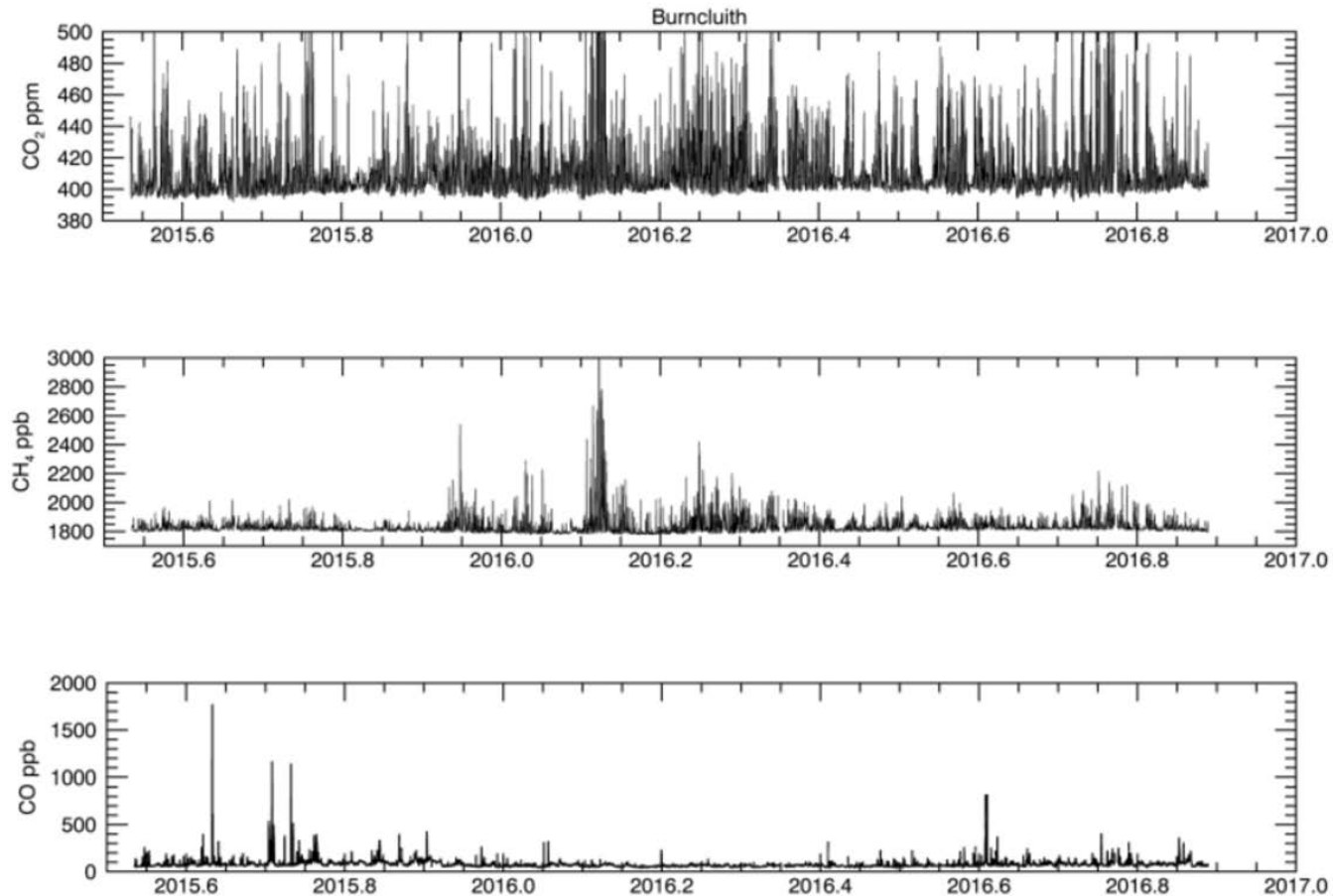


Figure 4. Measured concentration time series (hour means) of CO₂ (ppm), CH₄ (ppb) and CO (ppb) at Burncluith.

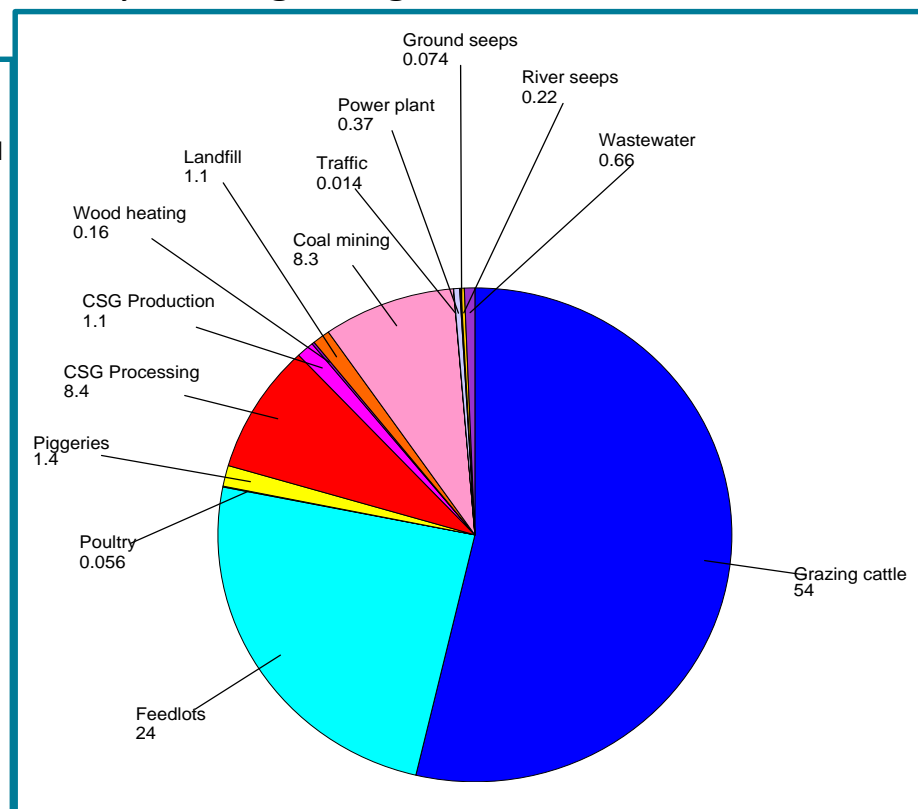
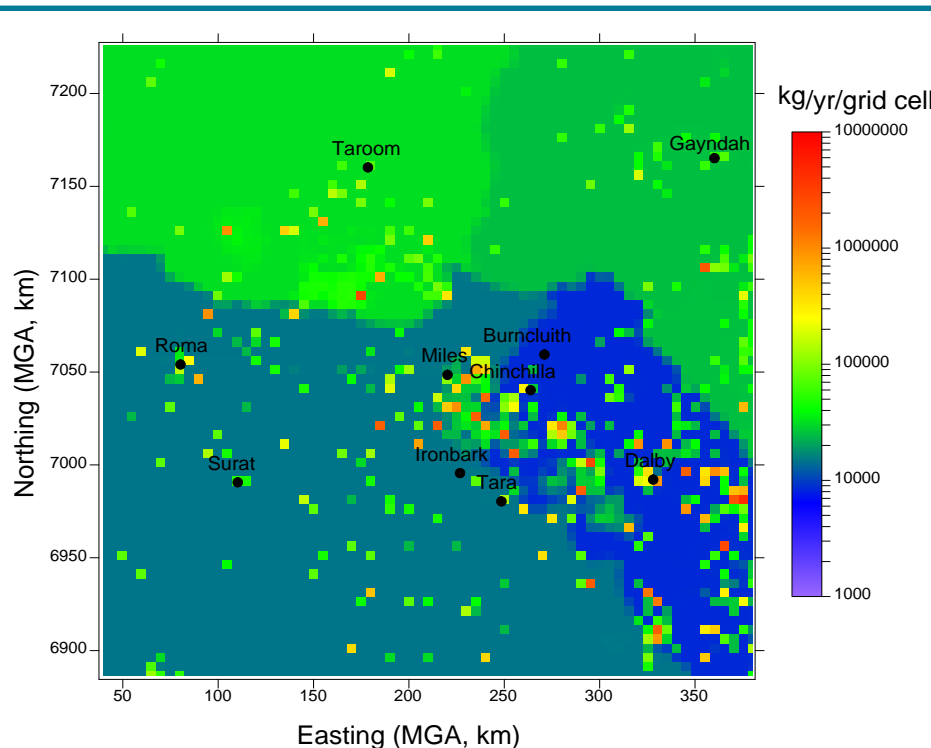
Data selection and filtering

Removal of unwanted signals

- Cows near analyser inlets (detected by CO₂)
- Burning off and dwelling open fire (CO)
- Nocturnal data (high stability, low wind speeds, extreme concentration gradients)

Bottom-up methane emission inventory for the region

- Prepared by Katestone Environmental with CSIRO input and feedback
- Used in forward runs and as a prior in the inverse modelling
- 1 km grid cells across 350 km x 350 km
- Total emission = $173 \times 10^6 \text{ kg yr}^{-1}$, dominated by cattle grazing, feedlots and CSG



CSG sources (Katestone inventory)

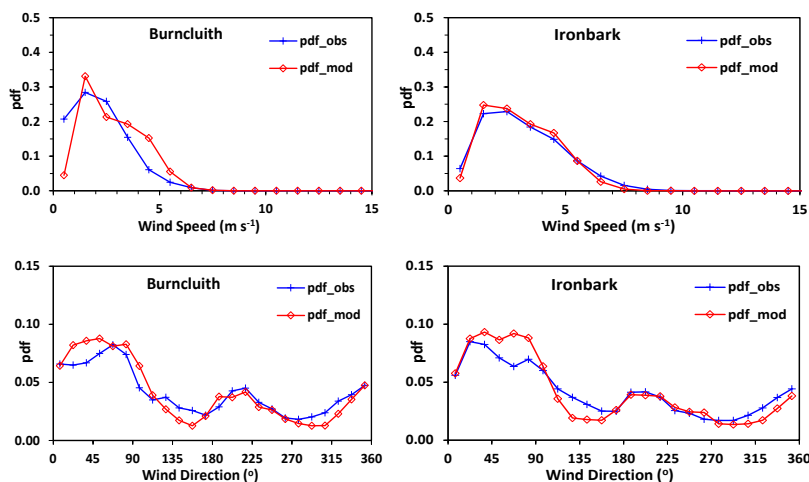
		Emission source	Intermittent	Continuous
Production emissions	Wellhead emissions	Wellhead control equipment		X
		Separators		X
		Maintenance	X	
		Leaks	X	
	Combustion emissions	Well head pumps		X
		Flaring	X	
		Diesel used in vehicles	X	
		Backup generators	X	
	Pipeline emissions	Pipeline control equipment		X
		High point vents on produced water pipelines	X	
Processing emissions	Processing facility emissions	Compressor venting		X
		Control equipment		X
		Gas conditioning units including dehydrators		X
	Combustion emissions	Plant compressors		X
		Flaring	X	
		Diesel used in vehicles	X	
		Backup generators	X	
	Produced water	Collection and storage of produced water	X	

Number of Operators	Number of Gas Fields	Number of Wells [^]	Number of Processing Facilities
Five	16	4628	16
Table note: [^] Number of wells estimated based on Queensland Government CSG production data			

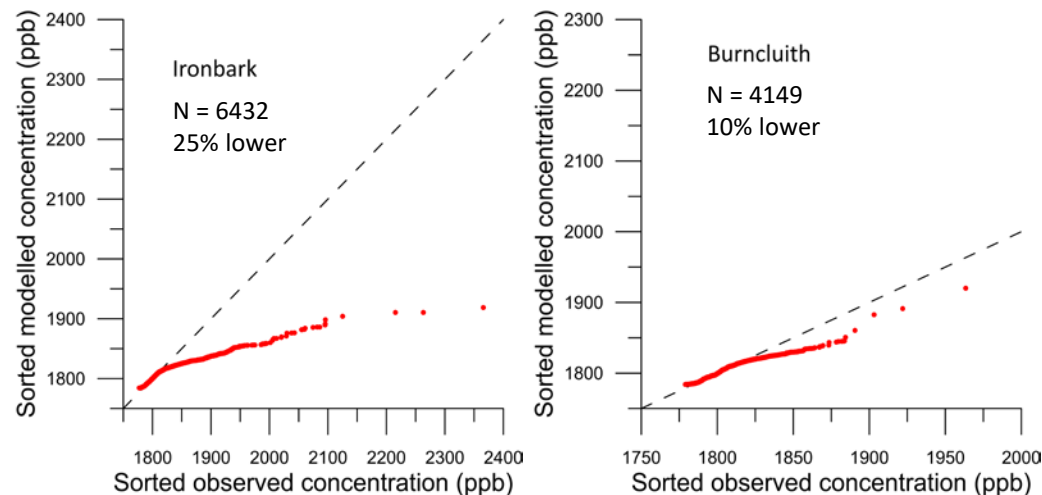
Forward modelling with bottom-up emissions

- CSIRO's forward prognostic model TAPM used
- The modelled meteorology compares well with observations
- Quantile-quantile (q-q) plots show that the model underestimates CH₄ observations suggesting missing sources or under-reported emissions in the inventory

Meteorology



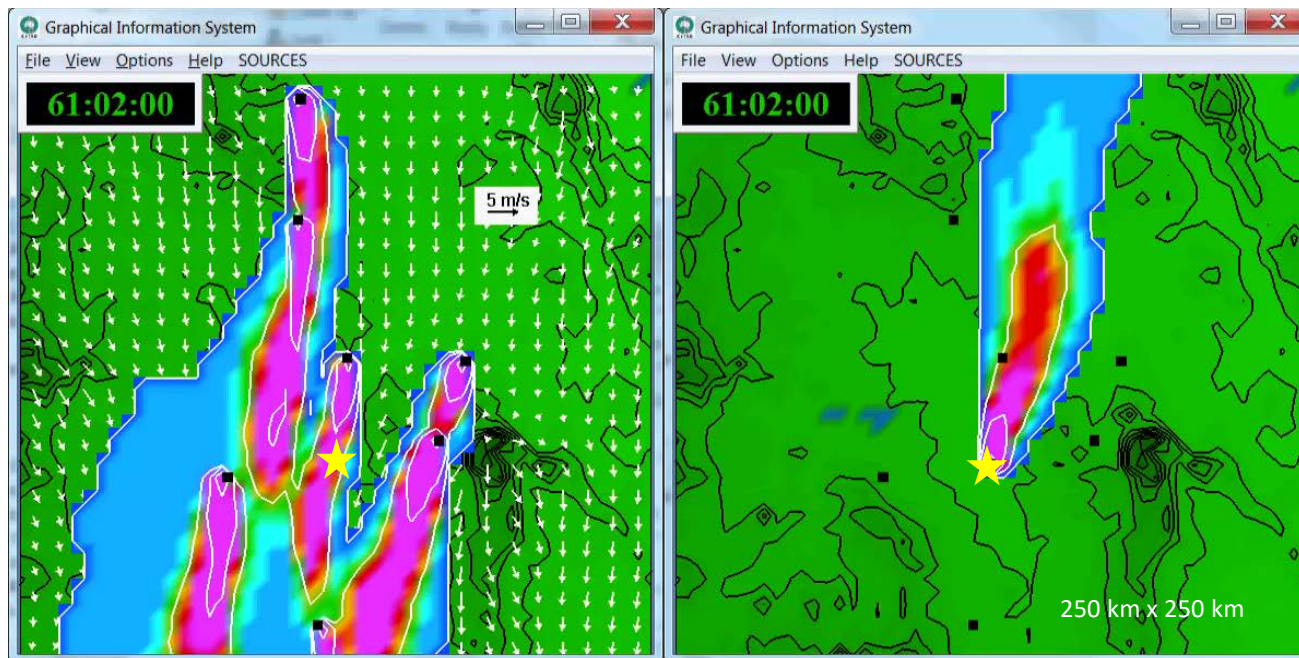
Methane concentration



Inverse modelling at local to regional scale

- TAPM formulated in backward mode for source-receptor relationship
-more efficient than forward
- Based on a Bayesian approach
- MCMC used for posterior sampling

Forward transport from sources Backward transport from receptor (monitor)



Bayes' rule

Posterior

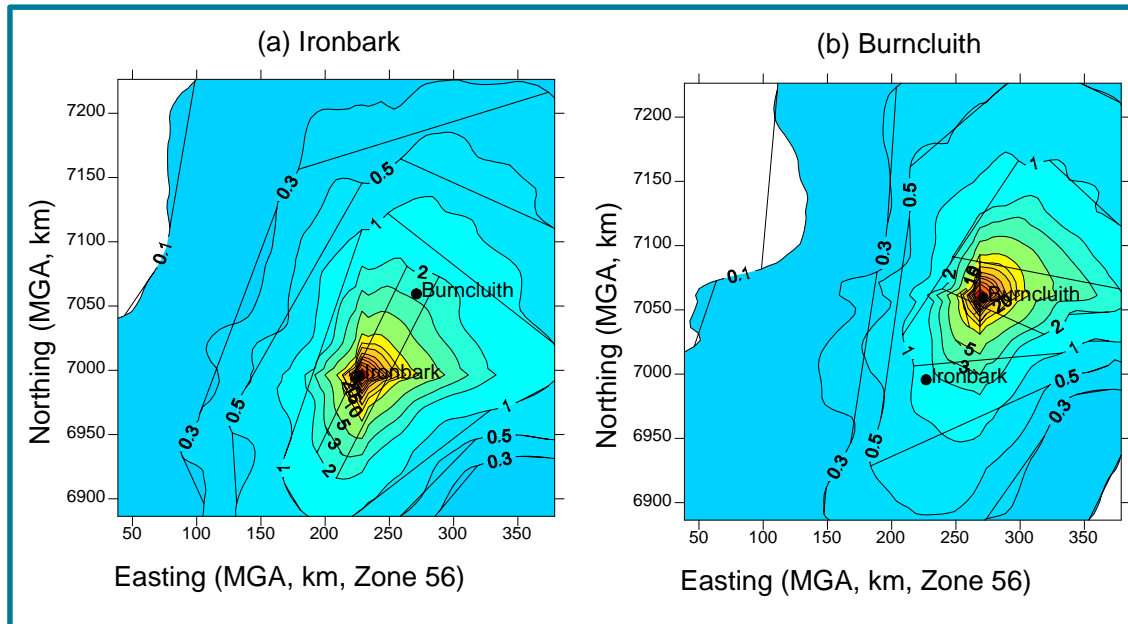
Prior

$$p(\mathbf{q} | \mathbf{c}) \propto p(\mathbf{q}) \cdot p(\mathbf{c} | \mathbf{q})$$

Likelihood function /
source-receptor relationship

Inverse model application for CH₄ emissions

- Tracers released from Ironbark and Burncluith (backward TAPM) to generate the source-receptor relationship required for the Bayesian analysis

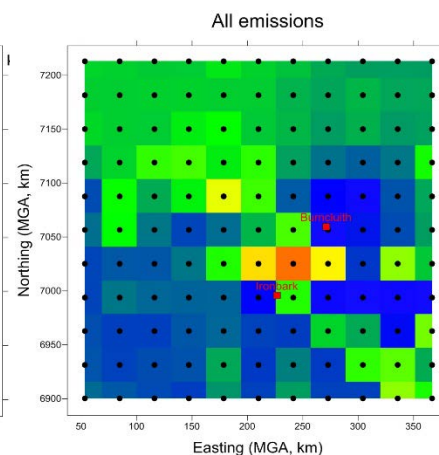
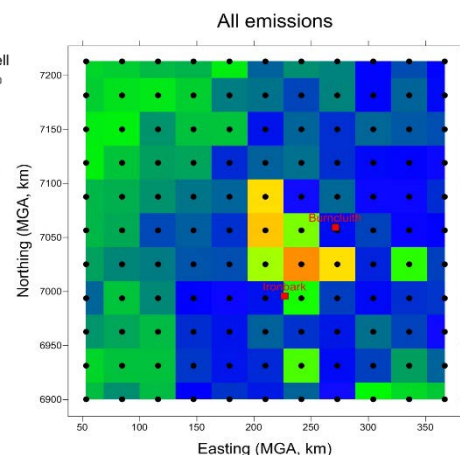
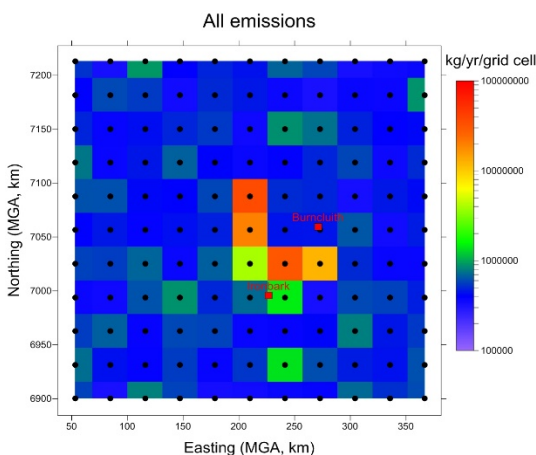


- Relatively low probability of adequately sampling the NW and SE corners of the domain
- Region of CSG activity between the two monitoring stations best sampled

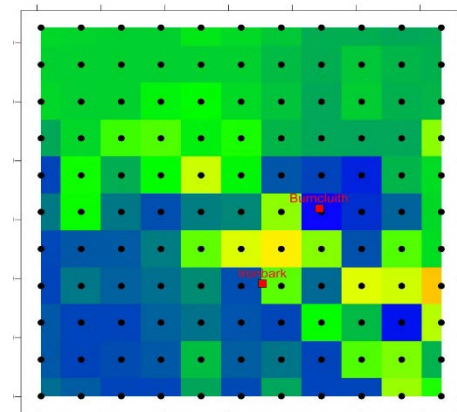
Simulation details

- 11 x 11 source regions considered (31 x 31 km)
- July 2015-December 2016
- Model and background methane uncertainties were accounted for
- Three cases of emission prior specified
 - 1) Loose bounds (10-10,000 g s⁻¹ per source area) – uninformative prior
 - 2) Spatially uniform prior (45.4 g s⁻¹ per source area), Gaussian uncertainty of 10%
 - 3) Bottom up inventory as prior, Gaussian uncertainty of 3%

Results: inferred emissions



Inventory, $173 \times 10^6 \text{ kg yr}^{-1}$



1) Uninformative prior

- Total emission within 6.4% of inventory
- High emissions in the centre consistent with inventory, but magnitude larger

2) Spatially Uniform prior

- Total emission within 17.7% of inventory
- Emissions distribution improved

3) Inventory as prior

- Total emission within 4.4% of the inventory
- $166 \times 10^6 \text{ kg yr}^{-1}$
- Distribution very similar to the prior but higher emissions between the two stations

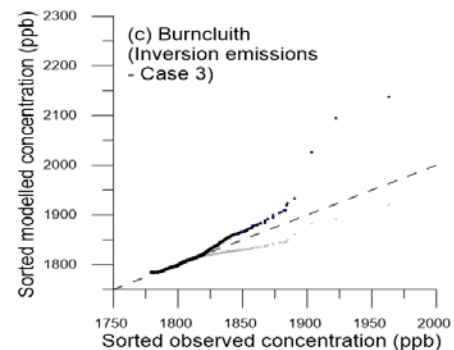
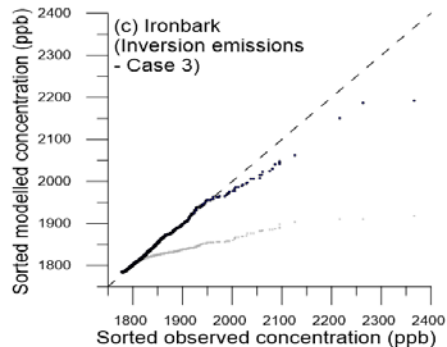
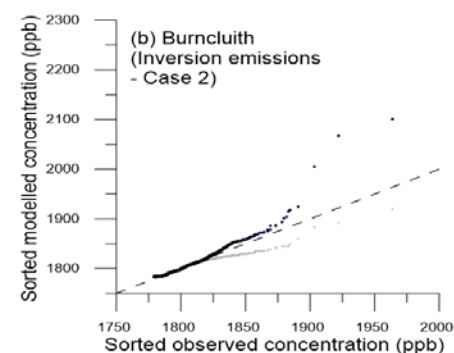
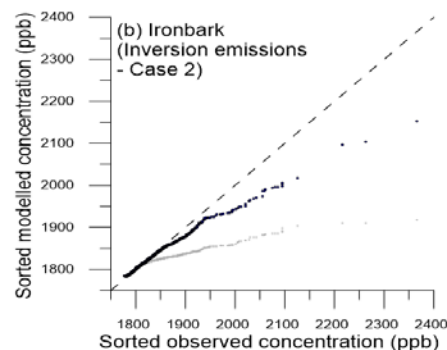
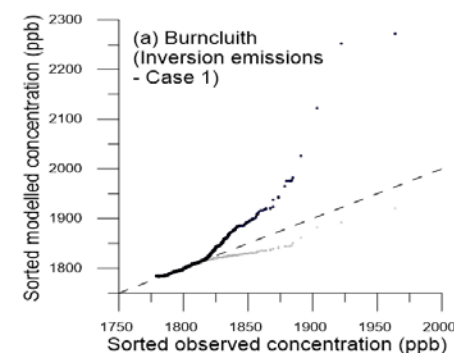
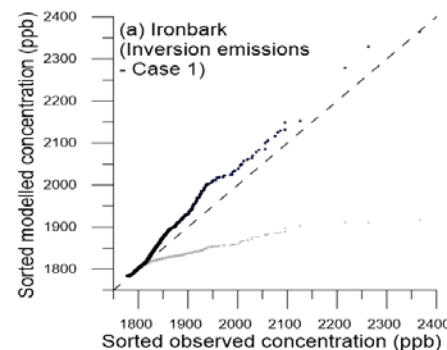
Inverse model validation

Faint symbols: with
inventory emissions

Inferred emissions are used in forward TAPM to simulate methane concentrations

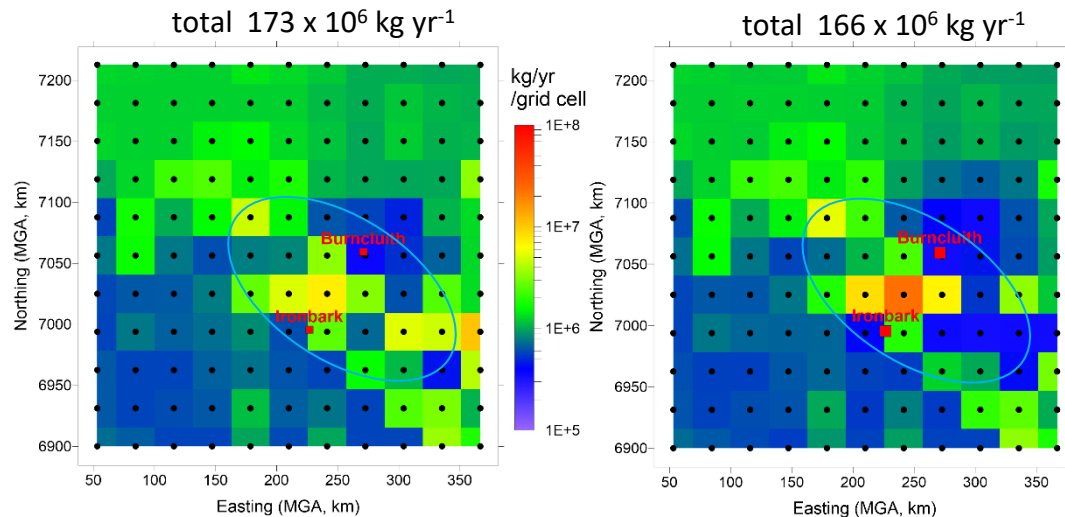
- Case 1: Loose bounds, uninformative prior
- Case 2: Spatially uniform prior
- Case 3: Bottom up inventory as prior

Case 3 provides the best comparison, but
Case 2 is not far off



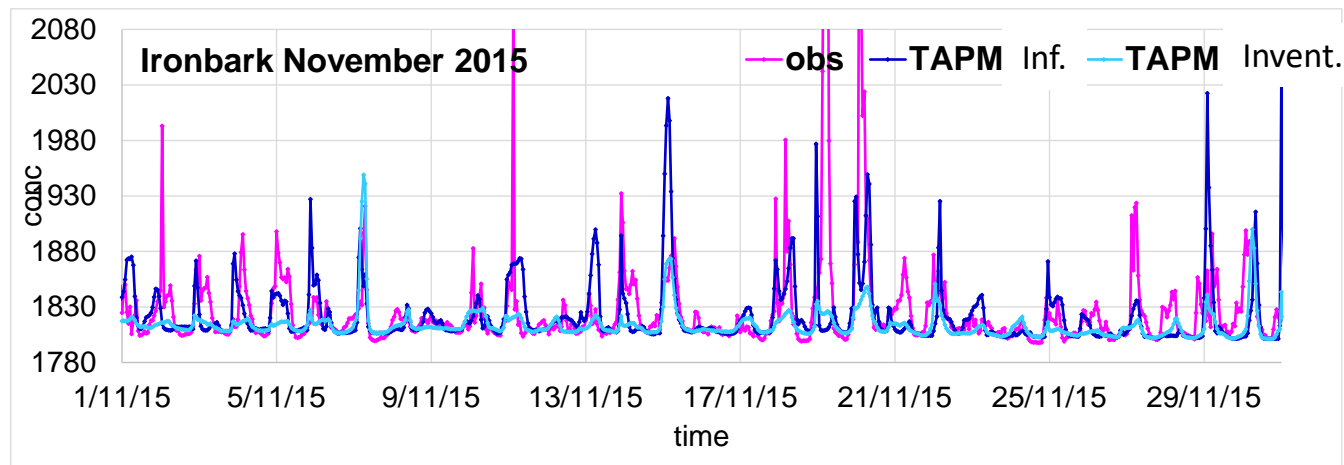
Emissions in CSG subregion

Bottom-up inventory emissions Case 3: Inversion inferred emissions



- Total inferred emissions similar to inventory, but 30% greater in the subregion
- Subregion dominated by feedlots + poultry + piggeries (30%), followed by cattle grazing (28%) and CSG processing (27%)

Observed and modelled timeseries



- The inferred emissions describe the observed concentrations (timing and size of peaks) better than the bottom-up emissions

Conclusions

An atmospheric “top down” methodology was developed to estimate CH₄ emissions from local to regional scale

- Combines a Bayesian inference approach and a backward configuration of TAPM
- Applied to the Surat Basin: 2 monitoring stations across 350x350 km
- Precise, inter-calibrated CH₄ concentrations, CO₂ and CO tracers, meteorology
- Stable solution, total emissions ($166 \times 10^6 \text{ kg yr}^{-1}$) and distributions compare well to prior information and bottom up inventory ($173 \times 10^6 \text{ kg yr}^{-1}$)
- In the CSG region, the inferred emissions are 30% greater than the inventory emissions
- Emissions inferred from inverse modelling explain the observed CH₄ concentrations better than the inventory
- Study described in full in Final Report and presented at three conferences including 2019 European Geophysical Union General Assembly

Further work

- Journal publication
- Explore value in other data – moving platforms (aircraft, vehicles), small low cost sensors, satellites
- Additional tracers to quantify source types – CH₄ isotopes, accompanying gases
- Follow up studies (after future growth and eventual wind down in CSG activity)
- Zone in on “hot spots” indicated by inversion

Acknowledgements



- CSIRO's Gas Industry Social and Environmental Research Alliance (GISERA)
Research reports <https://gisera.csiro.au/project/methane-seepage-in-the-surat-basin>
- CSIRO Oceans and Atmosphere (GASLAB), Energy, Land and Water, AIM Future Science Platform
- Katestone Environmental Pty Ltd (Natalie Shaw, Lisa Smith, Tania Haigh, Michael Burchill)
- CSG companies for activity data
- CSIRO Internal Reviewers (Mark Hibberd and Martin Cope)
- Land owners (G. and S. McConnachie; Origin Energy)



GISERA

Gas Industry Social and
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Thank you

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