

# Predicting demand for rehabilitation or GEM care from acute care data

ABF Conference,  
Brisbane, May 2016

# Aim of this project

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To build a national model that predicts demand  
...from acute inpatient activity  
...for subsequent rehabilitation and GEM care.  
...for initial application in Tas.

## **Model features:**

- Based on the concept of rehabilitation-sensitive DRGs
- Quantifies "sensitivity" rather than Yes/No
- Considers other relevant patient-level variables
- Concept of probability "tier" – high, medium, low or zero.

# In our toolbox we had....

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- Rehab-sensitive DRGs – Dr Lynette Lee.
- Some literature findings.
- National acute inpatient data (2 years) with subacute flags if <28 days in same hospital (2.6mill records).
- AROC benchmarking data.
- Clinical and technical expert advisors.
- Tasmanian inpatient data to test and apply the model.

# The rehabilitation-sensitive DRGs

177 AR-DRGs (Version 5.2), grouped into 17 “functional loss” groups:

- Amputee
- Arthritis
- Arthritis after care
- Chronic pain (back and neck)
- Complex joint replacement
- Complex medical
- Fractured neck of femur
- Joint replacement
- Multi trauma and other
- Neurological conditions
- Non-traumatic brain dysfunction
- Other complex orthopaedic
- Other orthopaedic
- Rehabilitation/other
- Spinal cord dysfunction
- Stroke
- Traumatic brain dysfunction

# Examples: rehab-sensitive DRGs

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- Stroke
  - B69A TIA and Precerebral Occlusion w Cat or Sev CC
  - B70B Stroke w Severe CC
  
- Traumatic brain dysfunction
  - B02A Craniotomy w Cat CC
  - B78B Intracranial Injury wo Cat or Sev CC
  - B79Z Skull Fractures
  - B80Z Other Head Injury

# What did the literature say?

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- Factors influencing referral for subacute care include;
  - Medical stability
  - Family and social supports
  - **Age**
  - Functional capacity
  - **Type of presenting condition**
  - **Acute onset of condition (especially injury)**
  - Cognition and willingness to participate.

# Why only adult inpatient Rehab and GEM?

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- There were too few inpatient episodes in the following groups to support statistical analysis;
  - Psychogeriatric
  - Palliative care
  - Paediatric subacute.

# Our approach

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- Iterative, with clinical advice and statistical analysis, each feeding the other.
- Clinical panel met initially to provide direction and at intervals throughout project.
- Wider clinical advice sought to test assumptions and findings of statistical analysis.



# What was the clinical advice?

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- This model approach was appropriate for predicting service demand for populations and should not be applied to individual patients.
- Agreed with the literature findings.
- The concept of ‘tiers’ of probability, low/ medium/ high (or none) was meaningful and useful.
- Patients with an inpatient LOS greater than 10 days would experience deconditioning and require some rehabilitation.

# Regression analysis

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To find the patient factors in acute episodes that best explain referrals to rehab or GEM care.

Would have been straight-forward if;

- Services were classified as rehabilitation or GEM consistently. Overlapping concepts of geriatric rehab and 'slow stream' rehab.
- We had national data with the acute care and subsequent rehabilitation or GEM episodes linked.
- All relevant data items were reliably collected in national data

.....but this was not the case....

# To counter these issues we;

- Combined the rehab and GEM data in the regression analysis and in the model output.
- Looked to other data sources to get some insight into what we were missing. This included AIHW publications and the Australasian Rehabilitation Outcomes Collaboration (AROC) data set.
- Used the variables that were available in the data set in the first instance and found that these had sufficient explanatory power.
  - The model may be refined later if the collection of patient risk factors such as family support is improved.

# How could we predict what the missing data would have looked like?

- The AROC and AIHW aggregate data gave us a measure of the prevalence we would have had if we'd had the ideal dataset.
- ADRGs were notionally mapped to the impairment groups within AROC data set.
- Statistical techniques then used to modify the probabilities in the model as if we had complete data.
- In selected ADRGs, based on clinical advice, the probabilities were further boosted by flagging acute episodes with LOS greater than 10 days.

# Building the model

- Logistic regression using the acute care data.
  - Significant variables - age and ADRG
- Probabilities identified by ADRG based on actuals.
- The 'tier' allocation assigned by clinicians also informed the model development and interpretation.
- **Model:**
  - A table of probabilities with ADRG as row and age group as column, and
  - A table of probabilities by MDC with tier as row and age group as column.

# MDC/tier example

**MDC 08** - Diseases and disorders of the musculoskeletal system and connective tissue

Tier	Age group				
	17-54	55-64	65-74	75-84	85+
low	0.004	0.013	0.024	0.063	0.102
medium	0.034	0.101	0.177	0.369	0.496
high	0.093	0.249	0.387	0.632	0.744

# MDC 08 ADRG examples

ADRG	Tier	Age group				
		17-54	55-64	65-74	75-84	85+
I17	low	0.002	0.007	0.013	0.032	0.050
I21	med	0.039	0.111	0.186	0.372	0.484
I31	high	0.134	0.326	0.470	0.696	0.784

# Key assumptions in developing the model:

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- It is appropriate to make predictions about the need for subacute care using acute episode of care data.
- The clinical/demographic profile of the flagged episodes in the acute data (60K) were an indicative cross section of patients who receive subacute care.
- It is clinically meaningful to combine rehabilitation and GEM in the model.



# Assumptions cont.

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- By using patient (rather than service) specific criteria we avoid having to account for different service delivery models.
- A mapping between the Adjacent DRG (ADRG) and rehabilitation impairment code in the AROC data is meaningful and appropriate.
- LOS in subacute care can be predicted meaningfully based on impairment code.

# Applying the model

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- Can apply to any aggregate data set that includes ADRG and age variables. Most useful to apply to a hospital cluster or LHD.
- Produces numbers of expected patient care episodes.
- Aggregate output by ADRG or MDC. Can group to;
  - Locally defined functional diagnosis groups for translation for service planning.
  - AROC (SNAP) impairment types (ie stroke, amputation, fractures etc).

# Translating the model output

- Used the notional map between ADRGs and impairment groups.
- AROC data provided average length of stay for impairment groups.
- This enabled us to predict the required bed days.
- Assuming (eg 90%) occupancy throughout the year, convert bed days to beds.
- *Required services may be set up as inpatient or inpatient equivalents in a non-admitted or community setting – response must be in the context of local service delivery arrangements.*

# A comparison between two populations

	Separations				Bed days			
	Provided		Predicted		Provided		Predicted	
Zone	Total	per 1,000 pop.	Total	per 1,000 pop.	Total	per 1,000 pop.	Total	per 1,000 pop.
Zone A	1,260	4.2	1,560	5.2	23,436	78	29,016	97
Zone B	560	2.8	1,300	6.5	11,256	56	26,130	131
Total	1,820	3.6	2,860	5.7	34,692	69	55,146	110

# What did the model analysis tell us?

- The utilisation of designated rehab and GEM services in Tas was significantly lower than other jurisdictions.
- Across Tas there was a requirement for an additional 114 rehab/GEM beds to meet the predicted demand.
- Most of this demand could be met by converting acute bed capacity to subacute beds (calculation based on selected short stay acute DRGs).
- The net deficit in rehab/GEM beds in Tas was only 13.
- Applied to the national data set the model also predicted 83 % of the actual total rehab and GEM activity.
- Powerful predictive tool.

# Does it pass the pub test?

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- It is big on assumptions!
- Was developed using data with significant limitations.

However...,

- It has been clinically validated and supported,
- It has produced meaningful and sensible results,
- Results can be interpreted in a local context.

# Acknowledgments

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- AHSRI team - Janette Green, Conrad Kobel, Habibur Seraji, Suanne Lawrence
- Clinicians – esp Lyn Lee, Chris Poulos
- Tasmanian DHHS reps

Thank you!

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