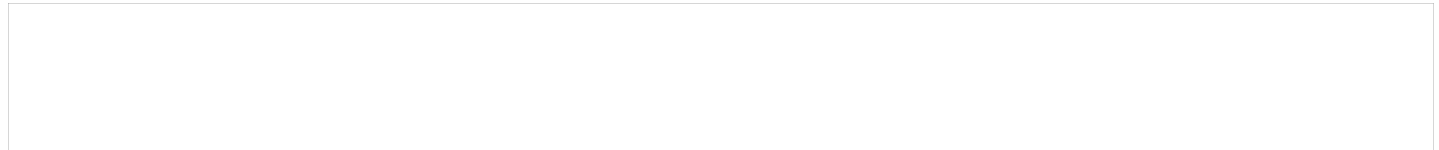
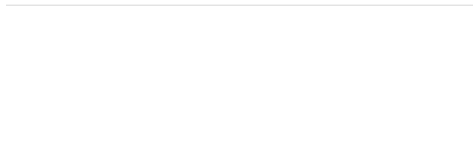


THE INCIDENCE AND COSTS OF CHEMOTHERAPY SIDE EFFECTS

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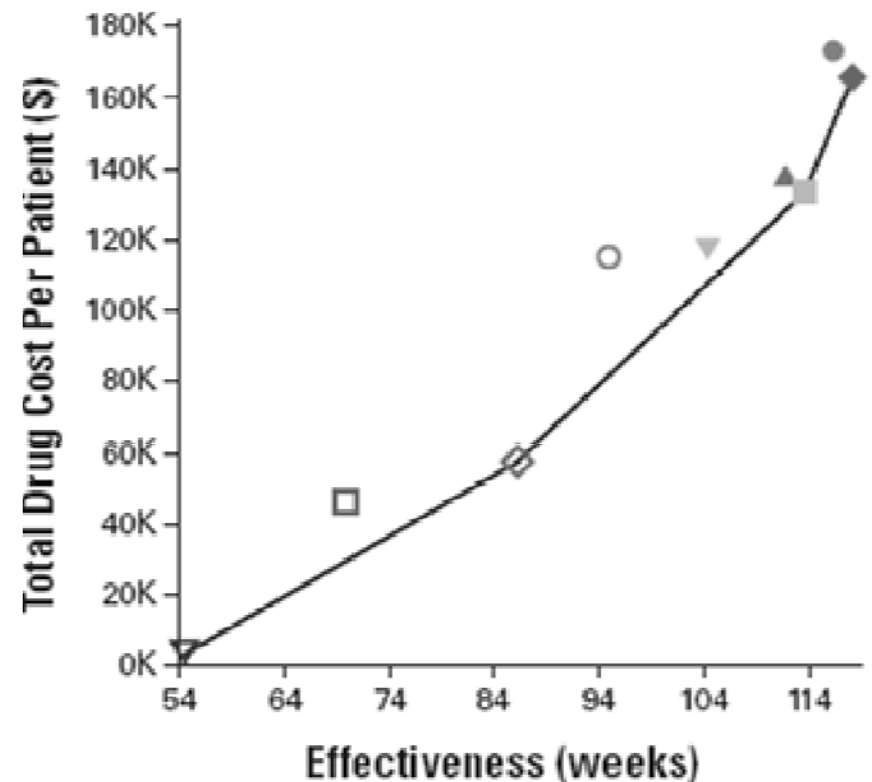
Supervisors: Marion Haas, Rosalie Viney

CAER 2013



Chemotherapy

- Chemotherapy drugs can be life extending for people with cancer. But...
 - ▣ they contribute a small amount to survival
 - ▣ they are increasingly expensive
 - ▣ they cause side effects



Chemotherapy side effects

- Chemotherapy side effects can:
 - ▣ Impact on patients physical wellbeing
 - ▣ Impact on patients quality of life (QoL)
 - ▣ Potentially impact on cancer survival
 - ▣ Be expensive to manage



Economic evaluation

- In Australia, new drugs are listed for public subsidy by PBAC on the basis of economic evaluation
- Literature review examined how side effects are incorporated into economic evaluations of chemotherapy
 - ▣ Costs and outcomes of side effects are not included in any systematic way
 - ▣ Clinical trials are the primary source of probabilities
 - ▣ Resource use is often estimated with expert opinion or based on best practice
- These data sources may not reflect clinical practice
- If side effects aren't accounted for (accurately) then outcomes of economic evaluations may be biased

Aims & Objective

- Overall objective:
 - To better inform models of chemotherapy cost effectiveness
- Aims:
 - Explore in clinical practice:
 1. the incidence of chemotherapy side effects
 2. the factors which influence the incidence of chemotherapy side effects
 3. the resource use associated with chemotherapy side effects

Department of Veterans Affairs

- The Australian Government Department of Veterans Affairs provides services to nearly 500,000 war veterans and their families in Australia
- Clients with a 'gold card' are entitled to the full range of services at DVA's expense
- DVA has actively encouraged the use of their data to undertake pharmacoepidemiological research

Data linkage

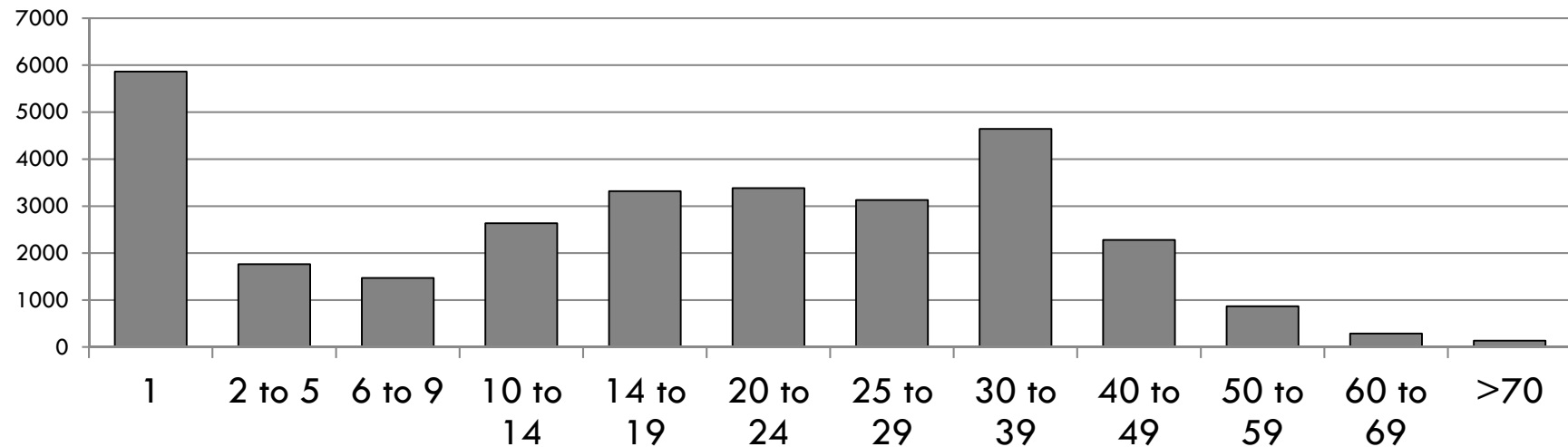
- Extract from DVA client database – individuals residing in NSW 1994 – 2007
- Linked by CHeReL to NSW population data

Registry	Start Date	End Date
NSW Cancer Registry	Jan 1994	Dec 2009
Repatriation PBS	01 July 2004	31 Jan 2010
Repatriation MBS	01 Jan 2000	31 Jan 2010
Admitted Patient Data Collection	01 July 2000	30 June 2009
Emergency Department Data	01 Jan 2005	31 Dec 2009
Resource utilisation period	01 Jan 2005	30 June 2009

Sample

□ Individual Gold Card Holders	1 29,307
▣ Individuals with a cancer diagnosis	29,480
▣ Individuals who received chemotherapy	12,030
▣ Total doses of chemotherapy	1 11,059

No. of PBS products per person with cancer



Demographics

Demographic	Chemo cohort
Proportion males	72%
Mean age (median) in years	81 (83)
age range	46 - 106
age group <70 yrs	14%
70-80 yrs	23%
>80 yrs	63%
Mean Rx Risk score (weighted comorbidities)	8.83
RxRisk score range	0 - 26

Cancer

Cancer site	N	% of cancer
Prostate	3124	39.17
Breast	1059	13.28
Melanoma of skin	881	11.05
Colon	491	6.16
Lung	354	4.44
Non-Hodgkin's lymphoma	349	4.38
Rectum, rectosigmoid, anus	279	3.5
Bladder	186	2.33
Ill-def & unspec site	136	1.71
Head & neck	591	0.65

Chemotherapy

Drug	Frequency	% of chemo	Used to treat...
Fluorouracil	2198	18.20	Breast, colorectal
Goserelin acetate	1909	15.80	Prostate, breast
Leuprorelin acetate	1307	10.82	Prostate
Bicalutamide	1005	8.32	Prostate, breast
Tamoxifen citrate	776	6.42	Breast
Capecitabine	327	2.71	Breast, colorectal
Rituximab	321	2.66	Lymphoma
Cyclophosphamide	305	2.53	Breast, leukemia
Anastrozole	280	2.32	Breast
Gemcitabine	276	2.28	Breast, lung, bladder, pancreas

Overview of methods

- 4 common side effects examined:
 - ▣ Diarrhoea, anaemia, nausea and vomiting (N&V), and neutropenia

- Aim 1 – incidence of side effects
 - ▣ The incidence of each side effect was calculated
- Aim 2 – factors influencing incidence of side effects
 - ▣ Multiple regression analysis using generalised estimating equations identified factors which influence the incidence of each side effect
- Aim 3 – resource use associated with side effects
 - ▣ Multiple linear regression identified whether those who experienced a side effect had higher chemotherapy costs

Overview of assumptions

- No direct data on whether someone experiences a side effect, so require a proxy
- Specific treatments are likely (based on best practice) to be given when an individual experiences a side effect
- These treatments can be related to chemotherapy administration by time
- In interpretation, need to consider:
 - ▣ “Individuals treated for a likely side effect”
 - ▣ individuals having these treatments for reasons other than side effects
 - ▣ individuals having side effects and not receiving these treatments
- Treatment of a side effect was considered related to chemotherapy when it occurred on or within three days after a chemotherapy dose

Incidence of side effects - method

- An analysis dataset was generated for each side effect
- For each dose of chemotherapy dispensed, a search was done of any side effect treatments which were given to the same individual within 3 days
- The incidence was calculated by dose of chemotherapy, and then by individual

Incidence of side effects - results

	Side effects	No. with chemotherapy	No. with side effect	% with side effect
By doses	Diarrhoea	89,594	879	1%
	Anaemia	84,872	638	<1%
	Nausea & vomiting	84,378	5,415	6%
	Neutropenia	84,495	601	<1%
By person	Diarrhoea	7,978	396	5%
	Anaemia	8,158	330	4%
	Nausea & vomiting	9,173	1,535	17%
	Neutropenia	8,069	242	3%

Factors influencing side effects - methods

- Multiple regression used to identify factors which influence the incidence of each side effect
- Binary outcome, so logistic model required
- Correlated data noted
 - Can restructure data to remove correlation, using a summary measure (eg: ever had a side effect), or
 - Can use technique designed for correlated data, such as Generalised Estimating Equations (GEE)

Generalised estimating equations

- Allow the correlation of outcomes within an individual to be estimated and taken into account in the regression coefficients and their standard errors
- The regression coefficients obtained from GEE are correctly interpreted in a population averaged manner
- Specifications of my GEE models
 - ▣ Repeated subject variable: PPN
 - ▣ Distribution: Binomial
 - ▣ Link function: Logit

GEE Correlation structures

- Independent – simplest assumption, but usually incorrect
 - ▣ Each observation for an individual is uncorrelated with every other observation for that individual.
 - ▣ The GEE reduces to the independence (GLM) estimating equation
- Exchangeable (compound symmetry)
 - ▣ Every observation within an individual is equally correlated with every other observation from that individual.
 - ▣ Fully characterised by the intraclass correlation coefficient
- Auto-regressive
 - ▣ Derived from time series analysis
 - ▣ Two observations taken close in time within an individual tend to be more highly correlated than two observations taken far apart in time from the same individual.
- Others, inc unstructured and user fixed – more complicated and situation specific

Factors influencing side effects - methods

side effect ~ α + *gender* + *age* + *RxRisk* + *chemo* + *cancer* + ε

Variable	Levels
Side effect	Yes / No
Gender	Male / Female
Age	Continuous, or <70 years 70 – 79 years >79 years
RxRisk (comorbidities)	Quartiles (0-7, 8-9, 10-12, 13-26)
Chemo	Consolidated to 8 levels based on ATC code
Cancer	Consolidated to 7 levels based on ICD classification

Factors influencing side effects - models

- Tested correlation structures to maximise model fit with all variables at least aggregated level
 - ▣ Autoregressive consistently chosen as most appropriate
 - ▣ Indicates that there is correlation based on time as well as individuals
- Tested models with aggregated variable levels for age (continuous vs 4 levels) and chemotherapy category (2 categorisations each with 8 levels)
 - ▣ Model 1 (continuous age and standard chemo categories) most appropriate for $\frac{3}{4}$ side effects

Summary of results

Variable	Diarrhoea	Nausea & vomiting	Anaemia	Neutropenia
Gender (female)	ND	Increase***	ND	ND
Age (younger)	Increase***	Increase***	ND	ND
RxRisk (fewer co-morbidities)	Decrease*	Decrease*	Decrease***	Decrease**
* <0.05, **<0.01, ***<0.001				

- Females are 1.6 times more likely to experience N&V
- Every additional year of age decreases odds of diarrhoea by 4% and decreases odds of N&V by 3%
- Moving from highest to lowest RxRisk reduces odds of a side effect by 25% (N&V) to 60% (neutropenia)

Summary of results

Variable	Diarrhoea	Nausea & vomiting	Anaemia	Neutropenia
Breast cancer	ND	Decrease*	ND	Increase***
Colorectal cancer	ND	ND	ND	Increase***
Genital cancer	ND	ND	ND	Increase***
Lung cancer	Decrease*	ND	ND	Increase***
Non-solid tumours	Decrease*	Decrease***	ND	Increase***
Other	ND	ND	ND	Increase***
<input type="checkbox"/> Compared to urinary cancer:				* <0.05, **<0.01, ***<0.001

- diarrhoea odds were 70% lower in lung and 60% lower in non-solid cancers
- N&V odds were reduced by nearly half in breast cancer and by over 60% in non-solid tumours
- The increase of odds of neutropenia was highest for non-solid tumours (50-fold) and lung cancers (20-fold)

Summary of results

Variable	Diarrhoea	N&V	Anaemia	Neutropenia
Antineoplastic	Decrease***	Increase***	ND	Increase*
Progestogens	ND	Increase*	ND	ND
LHRH agonists	Decrease***	Increase***	Decrease**	Increase***
Anti-estrogens	Decrease*	Increase***	ND	Increase***
Anti-androgens	Decrease**	Increase***	Decrease***	Increase*
Aromatase inhibitors	Decrease*	ND	Decrease*	ND
Immunostimulants	ND	ND	ND	Increase***

* <0.05, **<0.01, ***<0.001

□ Compared to immunosuppressants:

- Antineoplastics lower odds of diarrhoea by over 70%
- Anti-androgens increased odds of N&V by 13-fold
- AIs decrease odds of anaemia by 84%
- Immunostimulants increased odds of neutropenia by 700-fold

Resource use - methods

Total cost ~ α + gender + age + RxRisk + cancer + doses + any se + ϵ

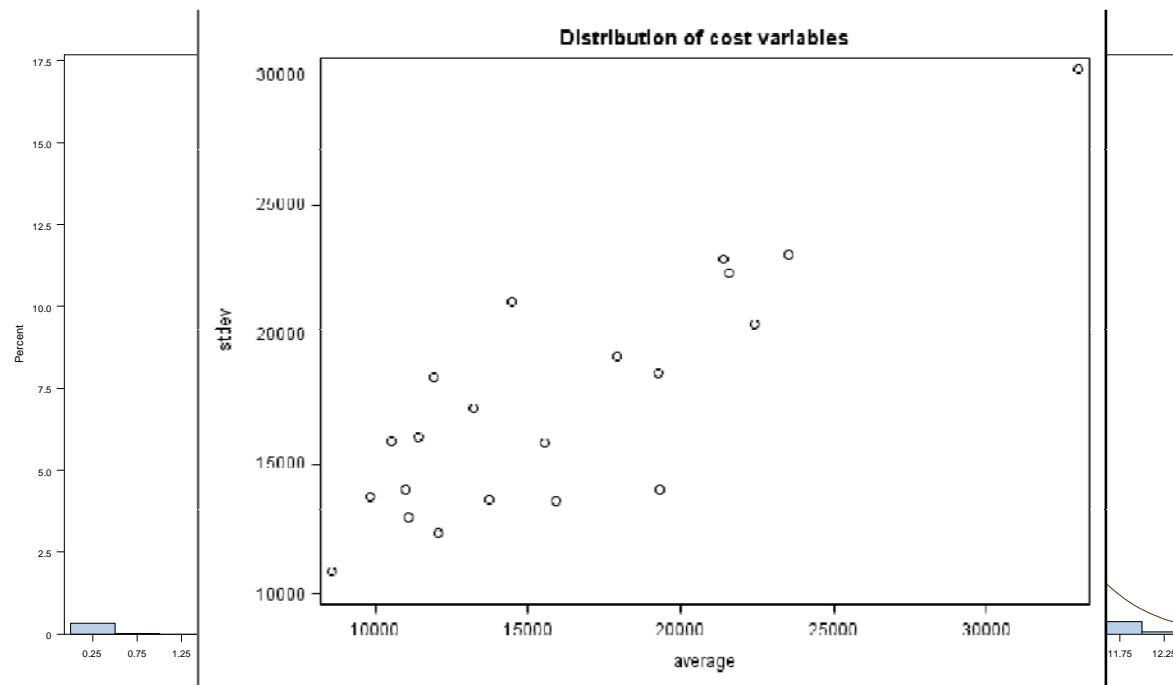
Variable	Levels
Total cost	Total health care expenditure (medical services, hospitalisation &/or pharmaceuticals) during the 6-month period following the first dose of a new chemotherapy regimen from 1 st Jan 2005
Gender	Male / Female
Age	<70 years 70 – 79 years >79 years
RxRisk	Quartiles (0-7, 8-9, 10-12, 13-26)
Doses	Total number of doses of chemotherapy (continuous)
Cancer	Consolidated to 7 levels based on ICD classification
Any side effect	Diarrhoea OR Anaemia OR N&V OR Neutropenia

Resource use- data distribution

- Cost data are typically positively skewed and truncated at zero, making parametric tests difficult
- Options include:
 - ▣ If large sample size, ignore skew (central limit theorem)
 - ▣ Non-parametric tests – inappropriate for decision makers
 - ▣ Transform data – retransformation difficult
 - ▣ Non-parametric bootstrapping – simulation method, but doesn't model the skewness of the data
 - ▣ Generalised linear modelling – allows responses to be distributed in other ways (often gamma distribution is appropriate for cost data)

Resource use - results

- Data highly skewed
- Log-transformed data approaches normal
- Mean vs standard deviation for raw costs shows an approximate constant coefficient of variation



Resource use – raw cost results

Solution for Fixed Effects - Simple linear regression of costs and each AE

Effect	Category	Estimate	Standard Error	Pr > t
Intercept		39705	3131.98	<.0001
Sex (vs male)	Female	-1418.69	599	0.0179
age		-140.26	30.3976	<.0001
RxRisk		552.77	59.6786	<.0001
Cancer site (vs urinary)	Breast	-4148.06	1299.15	0.0014
	CRC	616.02	1206.16	0.6096
	Genital	-3231.73	1097.67	0.0033
	Lung	237.14	1395.47	0.8651
	Non-solid	4655.44	1214.67	0.0001
	Other	-2693.62	1150.71	0.0193
Any diarrhoea	No	2498.68	977.5	0.0106
Any nausea/vomit	No	-7511.1	543.34	<.0001
Any anaemia	No	-4724.43	1042.62	<.0001
Any neutropenia	No	-10631	1141.47	<.0001

Resource use – log-transformed results

Solution for Fixed Effects - Regression of log costs – each AE

Effect	Category	Estimate	Standard Error	Pr > t
Intercept		10.2124	0.2335	<.0001
Sex (vs male)	Female	-0.2062	0.04466	<.0001
age		-0.00565	0.002266	0.0127
RxRisk		0.06941	0.004449	<.0001
Cancer site (vs urinary)	Breast	-0.3471	0.09686	0.0003
	CRC	-0.077	0.08992	0.3919
	Genital	-0.1911	0.08184	0.0195
	Lung	-0.167	0.104	0.1084
	Non-solid	0.1749	0.09056	0.0535
	Other	-0.3751	0.08579	<.0001
Any diarrhoea	No	-0.01491	0.07288	0.8379
Any nausea/vomit	No	-0.5665	0.04051	<.0001
Any anaemia	No	-0.3472	0.07773	<.0001
Any neutropenia	No	-0.5458	0.0851	<.0001

Resource use – GLM results

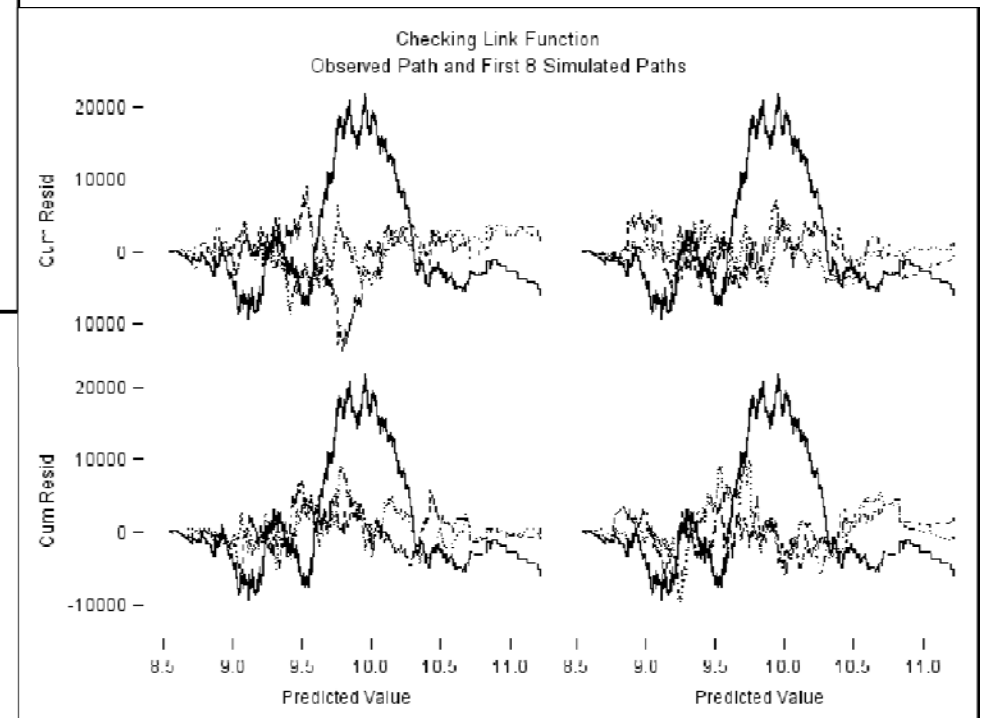
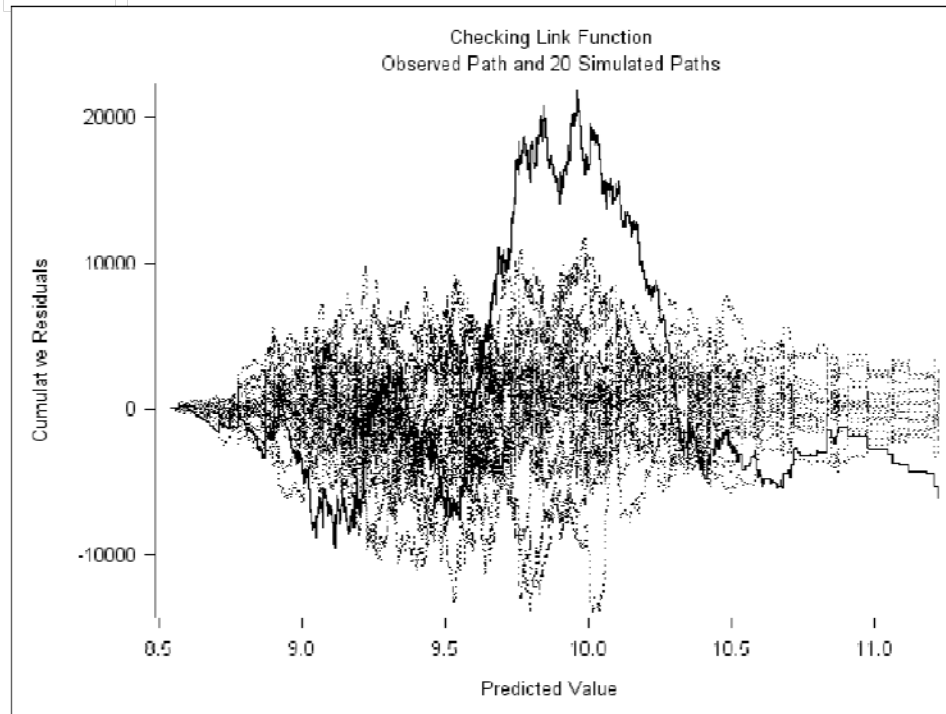
Parameter	Category	Exp (Estimate)	Exp (Wald 95% Confidence Limits)	
Intercept		14237.01	10515.44	19273.76
Sex	F	0.91	0.84	0.97
age		0.99	0.99	1.00
RxRisk		1.05	1.04	1.06
sitecatb	Breast	0.67	0.59	0.75
sitecatb	Genita	0.76	0.70	0.83
sitecatb	Lung	1.10	0.96	1.25
sitecatb	Nosoli	1.25	1.12	1.39
sitecatb	Other	0.76	0.69	0.83
sitecatb	Urinar	1.01	0.87	1.16
anydia	1	0.89	0.79	1.00
anynausea	1	1.61	1.51	1.72
anyanaemia	1	1.33	1.18	1.51
anyneut	1	1.54	1.34	1.76
Scale		2.95	2.85	3.06

Resource use – GLM results

□ Test model

- ▣ Plot cumulative residuals to assess fit of covariates or appropriateness of link function
- ▣ Assesses if the simulated residual patterns (with a log-link) that would be generated by the model under the specified assumptions are statistically different from the one actually generated

Resource use – GLM results



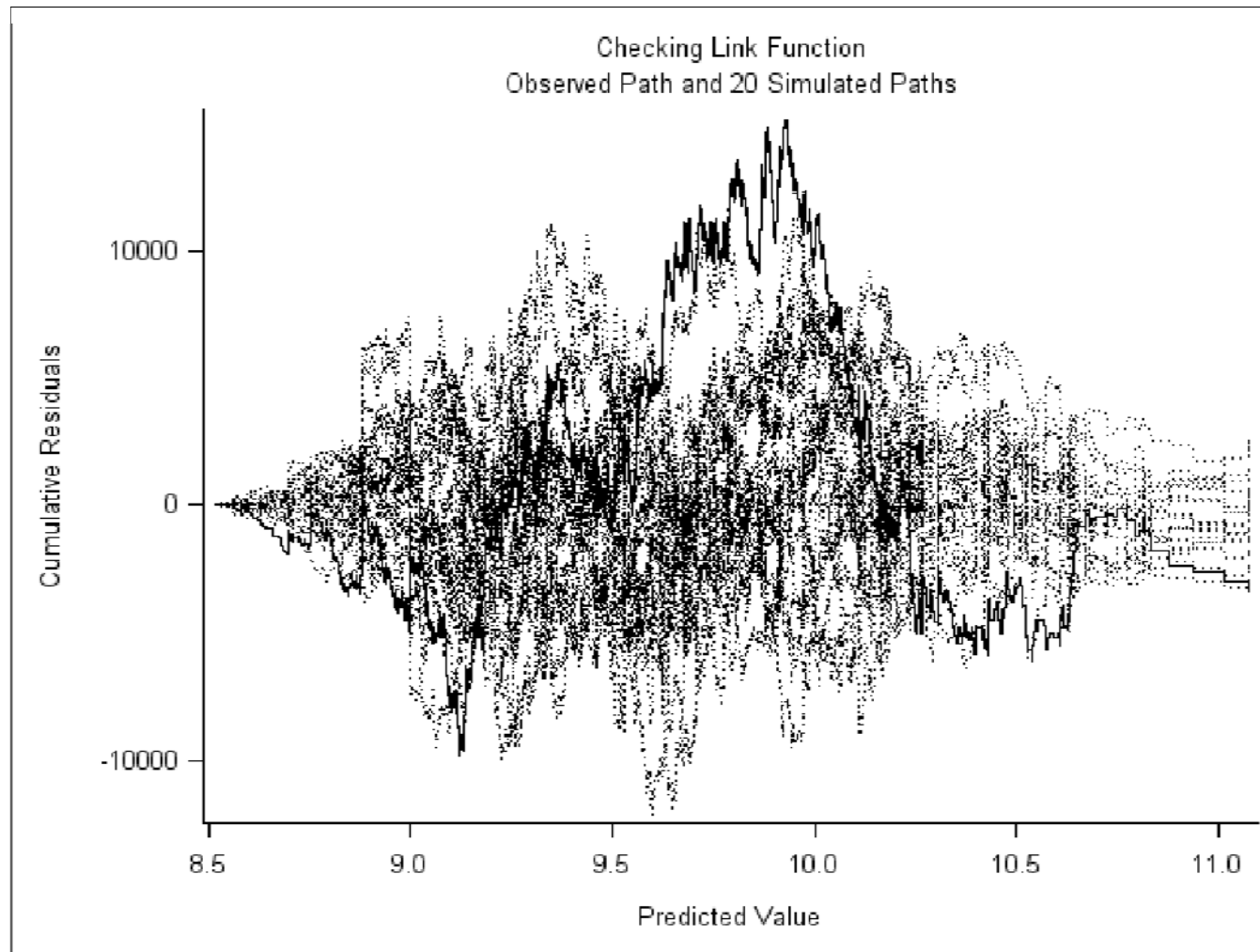
Resource use – GLM results

- Plots indicate an artefact in the data
- Exploratory analysis of model with interaction terms
 - Between side effects
 - 2/3 anaemia interactions were significant at $p < 0.05$ level
 - Between the type of cancer and the side effect
 - Nausea had the strongest association with type of cancer
 - Between age and comorbidities
 - Not significant

Resource use – GLM results

- Final model included:
 - ▣ Main effects
 - ▣ Interaction term for anaemia and other side effects
 - ▣ Interaction term for nausea and cancer type
- Little impact on the significance of the main effects on total cost
- A number of interaction terms appear to significantly influence total cost
- Inclusion of interaction terms appears to improve model fit

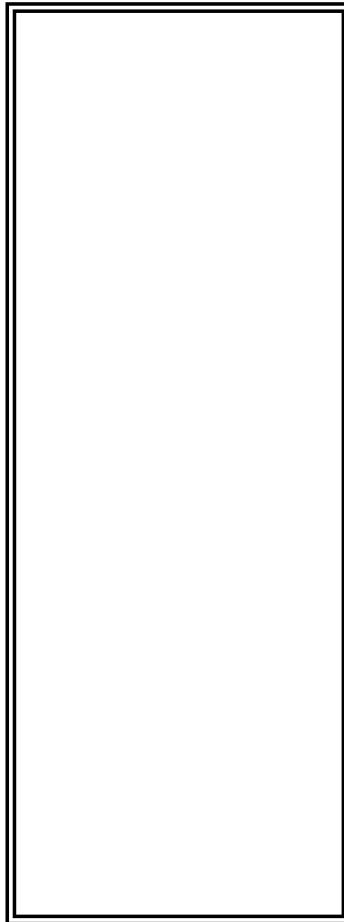
Resource use – GLM results



Conclusions

- This large administrative dataset provides an opportunity to examine ‘real life’ incidence of chemotherapy side effects in older people
- Being treated for a likely side effect is more common in individuals who are older or who have more co-morbidities
- Being treated for a likely side effect may be influenced by the type of cancer and chemotherapy an individual has
- Being treated for a likely side effect significantly increases overall healthcare costs

Acknowledgements



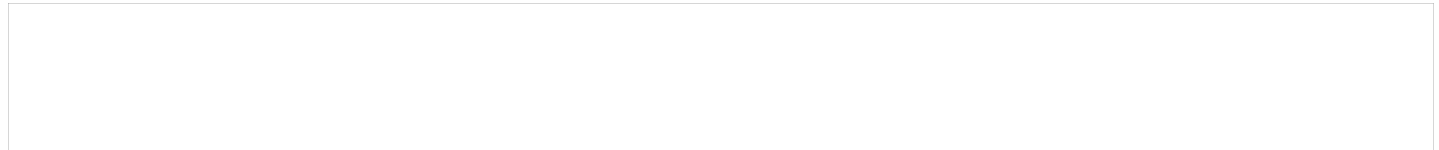
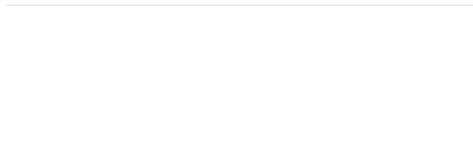
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THE INCIDENCE AND COSTS OF CHEMOTHERAPY SIDE EFFECTS

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Research and Evaluation, UTS

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CAER 2013



Spare slides

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Data issues – dataset size

- Large number of individuals, each with multiple observations
- 63Gb dataset for PBS data alone
- Data management and analysis techniques to manage this included:
 - ▣ Network location to reduce processing time
 - ▣ Removal of text variables to reduce data size
 - ▣ Use of SQL programming to improve efficiency

GEE Parameter Estimates (Odds Ratios)				
Parameter		Diarrhoea	N&V	Anaemia
Intercept		0.04***	0.06***	0.03***
Gender	Female	1.00	1.41***	0.81
	Male	.	.	.
Age	< 70 yrs	0.74	0.58***	0.26***
	70 - 80 yrs	1.35*	1.08	0.94
	> 80 Yrs	.	.	.
Rx Risk	0 to 7	0.57***	0.81*	0.49***
	8 to 9	0.70	1.09	0.77
	10 to 12	0.79	1.20	0.81
	13 +	.	.	.
Cancer	Breast	0.54	0.83	0.69
	Colorectal	2.38*	1.39	0.95
	Genital	0.64	0.99	1.18
	Lung	0.39*	2.49***	0.86
	Non-solid	0.29***	0.82	1.96
	Other	0.77	0.82	0.69
	Urinary	.	.	.
Chemo	Antineoplastic	0.58	2.87***	0.67
	Progestogens	0.49	1.96	0.41
	LHRH agonist	0.25***	0.28***	0.29*
	Anti-estrogens	0.47	0.40***	0.60
	Anti-androgens	0.33*	0.46***	0.26*
	Aromatase inhibitors	0.45	0.66	0.18
	Immunostimulants	0.53	1.20	1.49
	Immunosuppresants	.	.	.

Comparing correlation structures

GEE Model Information			
Correlation Structure	Exchangeable	Independent	AR(1)
Subject Effect	ppn (7976 levels)	ppn (7976 levels)	ppn (7976 levels)
Number of Clusters	7976	7976	7976
Clusters With Missing Values	5	5	5
Correlation Matrix Dimension	151	151	151
Maximum Cluster Size	151	151	151
Minimum Cluster Size	1	1	1
Algorithm converged	Yes	Yes	Yes
GEE Fit Criteria			
QIC	9450.3736	9433.6513	9429.4868
QICu	9378.4911	9320.2626	9348.8765
Exchangeable Working Correlation			
Correlation	0.151245715	.	.

Generalised Estimating Equations

- A GEE model extends generalised linear models in 2 ways:
 - ▣ Allows the correlation of outcomes within an individual to be estimated and taken into account in the regression coefficients and their standard errors
 - ▣ Permits the calculation of *robust* estimates for the standard errors (empirical standard errors) of the regression coefficients
 - ensures consistent inferences even if the correlation structure is incorrect, or the strength of the correlation varies between people
- *“Main benefit is that besides the seeking of more efficient estimators of regression parameters, the GEE produces reasonably accurate standard errors, hence confidence intervals with the correct coverage rates.”*

GEE continued

- GEE can deal with different number of observations per person (ie no need for designs to be balanced)
- GEE ideal when it is the fixed regression parameters which are of primary interest, and the correlation structure is merely a nuisance
- GEE not suitable if it is the correlation structure that you wanted to model (better to use multi-level modelling)
- Typically used in epidemiology and health, with responses that are binomially distributed or Poisson distributed.

Generalised Estimating Equations

- Allow the correlation of outcomes within an individual to be estimated and taken into account in the regression coefficients and their standard errors
- The correlation structure is pre-specified
- Despite incorrect working correlation structure the resulting regression coefficient estimate is still consistent
- Can deal with different number of observations per person
- Produces reasonably accurate standard errors, and therefore confidence intervals with the correct coverage rates

Steps to a GEE

1. Fit a standard regression model - assuming all observations are independent
2. Take the residuals from the regression and use these to estimate the parameters which quantify the correlation between observations in the same individual
3. Refit the regression model using a modified algorithm incorporating a matrix which reflects the magnitude of the correlation estimated in step 2
4. Keep alternating between steps 2 and 3 until the estimates all stabilize (ie model converges)

Interpreting GEE results

- The regression coefficients obtained from GEE are correctly interpreted in a population averaged manner
 - ▣ This is same as regular logit (specified for cluster)
 - ▣ I.e: Estimates the odds of the average male getting diarrhoea compared to the odds of the average female getting diarrhoea.
 - ▣ Different to random effects logit, which calculates the individual effect, ie the odds of a person having diarrhoea if male compared to the odds of the same person getting diarrhoea if female
- In many cases the population averaged and subject specific estimates are very close, but not always

GEE References

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- Hanley, Negass, Edwards, Forrester. Statistical analysis of correlated data using generalised estimating equations: An orientation. *American journal of epidemiology*. 2003, 157(4):364-375
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- Liang & Zeger. Longitudinal data analysis using generalised linear models. *Biometrics*. 1986, 78(1):13-22

Change data structure

- Summary statistics can be used to address the correlation
 - ▣ Changed data structure so only one observation per patient
 - ▣ Summary statistic 'ever diarrhoea' (assoc with chemo)
 - ▣ Replace chemotherapy details with number of doses of chemotherapy
 - ▣ Use simple logistic regression
- Removes potential patient correlation
- Similar analysis to that used in clinical trials
- Model converges

Logistic regression with summary statistic – model fit

Model Fit Statistics		
Criterion	Intercept	Intercept
	Only	and
		Covariates
AIC	3152.002	2862.676
SC	3158.986	2960.451
-2 Log L	3150.002	2834.676

Testing Global Null Hypothesis: BETA=0			
Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	315.3261	13	<.0001
Score	411.3727	13	<.0001
Wald	323.4319	13	<.0001

Logistic regression with summary statistic

Odds Ratio Estimates			
Effect	Point Estimate	95% Wald Confidence Limits	
Sex Female vs Male	1.137	0.852	1.519
Age <70 vs 80+	1.271	0.962	1.678
Age 70 – 79 vs 80+	1.792*	1.387	2.316
RxRisk 0-7 vs 13+	0.439*	0.325	0.595
RxRisk 8-9 vs 13+	0.723*	0.532	0.981
RxRisk 10-12 vs 13+	0.795	0.599	1.054
Cancer site Breast vs Urinary	0.719	0.361	1.433
Cancer site CRC vs Urinary	2.979*	1.605	5.526
Cancer site Genital vs Urinary	0.679	0.365	1.263
Cancer site Lung vs Urinary	0.627	0.275	1.428
Cancer site Nosoli vs Urinary	0.678	0.34	1.35
Cancer site Other vs Urinary	0.817	0.426	1.567
Chemotherapy doses	1.035*	1.03	1.04

Logistic regression with a summary statistic

- While this method has advantages, it is not always ideal
- It requires use of a summary statistic, which may not answer the question
- It is inefficient, as it only uses some of the data in the dataset
- Generalised Estimating Equations can be used when a simple logistic regression would be appropriate, but there is correlation in the data, such as repeated measures

GEE Parameter Estimates - Nausea & Vomiting

Parameter		Estimate (Odds Ratio)	Standard Error	95% Confidence Limits		Pr > Z
Intercept		0.06***	1.32	0.03	0.10	<.0001
Gender	Female	1.41***	1.09	1.19	1.67	<.0001
	Male
Age	< 70 yrs	0.58***	1.11	0.47	0.71	<.0001
	70 - 80 yrs	1.08	1.07	0.95	1.22	0.23
	> 80 Yrs
Rx Risk	0 to 7	0.81*	1.11	0.66	0.99	0.04
	8 to 9	1.09	1.11	0.89	1.34	0.40
	10 to 12	1.20	1.10	0.99	1.45	0.07
	13 +
Cancer	Breast	0.83	1.31	0.49	1.40	0.48
	Colorectal	1.39	1.25	0.90	2.16	0.14
	Genital	0.99	1.26	0.63	1.55	0.96
	Lung	2.49***	1.26	1.59	3.91	<.0001
	Non-solid	0.82	1.25	0.53	1.28	0.39
	Other	0.82	1.26	0.53	1.29	0.39
	Urinary
Chemo	Antineoplastic	2.87***	1.19	2.04	4.02	<.0001
	Progestogens	1.96	1.45	0.95	4.06	0.07
	LHRH agonists	0.28***	1.23	0.19	0.42	<.0001
	Anti-estrogens	0.40***	1.29	0.24	0.66	0.00
	Anti-androgens	0.46***	1.26	0.29	0.72	0.00
	Aromatase inhibitors	0.66	1.30	0.39	1.11	0.11
	Immunostimulants	1.20	1.47	0.56	2.56	0.64
	Immunosuppresants

GEE Parameter Estimates - Anaemia

Parameter		Estimate (Odds Ratio)	Standard Error	95% Confidence Limits		Pr > Z
Intercept		0.03***	1.85	0.01	0.09	<.0001
Gender	Female	0.81	1.23	0.54	1.21	0.30
	Male
Age	< 70 yrs	0.26***	1.41	0.13	0.51	<.0001
	70 - 80 yrs	0.94	1.16	0.70	1.26	0.68
	> 80 Yrs
Rx Risk	0 to 7	0.49***	1.22	0.33	0.72	0.00
	8 to 9	0.77	1.22	0.53	1.13	0.18
	10 to 12	0.81	1.20	0.56	1.15	0.24
	13 +
Cancer	Breast	0.69	2.10	0.16	2.94	0.61
	Colorectal	0.95	1.58	0.39	2.33	0.91
	Genital	1.18	1.55	0.50	2.80	0.71
	Lung	0.86	1.61	0.34	2.19	0.76
	Non-solid	1.96	1.52	0.86	4.45	0.11
	Other	0.69	1.58	0.28	1.69	0.42
	Urinary
Chemo	Antineoplastic	0.67	1.66	0.25	1.81	0.43
	Progestogens	0.41	2.36	0.08	2.19	0.30
	LHRH agonists	0.29*	1.70	0.10	0.81	0.02
	Anti-estrogens	0.60	2.04	0.15	2.42	0.47
	Anti-androgens	0.26*	1.72	0.09	0.75	0.01
	Aromatase inhibitors	0.18	2.71	0.03	1.27	0.09
	Immunostimulants	1.49	1.80	0.47	4.74	0.50
	Immunosuppresants

Interpreting results – diarrhoea

- No difference in the odds of the average male getting diarrhoea compared to the odds of the average female getting diarrhoea
- Moving from the highest RxRisk category to the lowest, the average person will experience a 43% reduction in the odds of getting diarrhoea
- Compared to urinary cancer, the average person with CRC is 2.4 times more likely to experience diarrhoea.
- For the average person, the use of LHRH agonists or anti-androgens will reduce the odds of diarrhoea, compared to immunosuppressant's

Interpreting results – Nausea & vomiting

- The average female is 1.4 times more likely to experience N&V than the average male
- The average individual in the younger age group has a 40% lower odds of experiencing N&V
- Compared to urinary cancer, the average person with lung cancer is more likely to experience N&V
- The average person on antineoplastic chemotherapy is almost 3 times more likely to experience N&V than the average person on immunosuppressants
- LHRH agonists, anti-estrogens and anti-androgens all decrease the odds of N&V in the average person

Interpreting results - anaemia

- No difference based on gender
- The younger age group has an 80% reduction in the odds of having anaemia
- Moving from the highest RxRisk category to the lowest, the average person will experience a 50% reduction in the odds of getting diarrhoea
- Type of cancer does not appear to have a significant impact on the experience of anaemia
- LHRH-agonists and anti-androgens reduce the odds of anaemia compared to immunosuppressant's

What next?

- Model impact of having one AE on odds of having additional AE(s)
 - ▣ Correlation analysis
 - ▣ Two-part model
- Incorporate results into decision models and examine differences in outcomes
- Questions?