Probabilistic and Deterministic Data Linkage between Kentucky Cancer Registry and Health Claims Data

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Background

- > Due to limited resource, cancer registry data do not capture complete treatment information.
- ➤ Report from IOM (Cancer Patient to Cancer Survival: Lost in Transition) recommended great focus on linking data between registry and administrative data.
- Provide empirical foundation for better studied to access quality of care and outcome.
- ➤ A lot registries have done such data linkage (GA, OH, NY, WV,...)





Background – Data Linkage Methods

Deterministic matching

- > Exact matching on key variables
- Matching is defined by a predetermined algorithm
- No consideration on how likely values are to agree by chance

Probabilistic matching

- Linkage scores based on properties of fields being matched
- M-probability and U-probability
- Linkage score depends on probability that two records match
- > A cutoff value will be used to define potential true matches.
- Involve manual review





Background – Previous Studies

- Previous studies
 - WV: Deterministic data linkage for Medicaid and Medicare data
 - NY: Probabilistic data linkage for Medicaid data.
 - OH: Deterministic data linkage for Medicaid and Medicare data
 - ➤ GA: Deterministic data linkage for Medicaid and commercial claims data
- ➤ The characteristics of probabilistic and deterministic data linkage have not been thoroughly examined between registry data and claims data.
- ➤ Kentucky Cancer Registry (KCR) is conducting a study to link registry data and health claims data, such as Medicare, Medicaid and private insurance group claims data. This provides opportunity to examine the characteristics of two data linkage approaches.





CDC Registry Plus Link Plus

- Many probabilistic data linkage software packages are available.
- Registry Plus Link Plus is free and developed for data linkage between registry data and data from other sources.
- Link Plus Version 3.0 (http://www.cdc.gov/cancer/npcr/tools/registryplus/lp tech info.htm)
 - ➤ Handle very large data sets (over 4 millions records for file1 and no limit on file 2)
 - Provide one-to-many matching and many-to-many matching
 - Nice manual review interface





Data Involved in the Linkage

- KCR and Medicaid data linkage
 - KCR: 2011 KCR registry data for six cancer sites (10,887)
 - Medicaid: 2011 Medicaid enrollment data, including SSN, name, birth date, gender, county code. Race is not reliable. (1,051,987)
 - > KCR has the access of the Medicaid enrollment data.
- KCR and Humana Data linkage
 - ➤ KCR: 2011 KCR registry data for six cancer sites
 - ➤ Humana: 2007-2012 Humana Enrollment data, including SSN, name, birth date, gender, specific address. Race is not reliable.
 - ➤ KY 1,714,465, IN 761,051, OH 1,339,385,
 - ➤ Rest neighboring states 3,696,381
 - > KCR doesn't have the access of the Humana enrollment data, which make the linkage process very challenging.





Process for Probabilistic Data Linkage

Linking criteria:

- Blocking variables: SSN, birthdate, first name, last name, middle name
- Matching variables: SSN, birthdate, first name, last name, middle name and sex
- ➤ Direct method: calculate M and U-probabilities from the distribution of file 1.
- Cutoff values: lowest linkage score to be considered as a match (5)
- > 1-to-Many: Records in File 1 can match multiple records in File 2.
- ➤ Many-to-Many: Records in both File 1 and File 2 can match multiple records.
- Manually review the potential matches to identify true matches





Link Plus – Screen Shot

# [Z:\Data Linkage Project\Year 1\Testing Kun\Medicaid\Medicaid_Testing, File Manual Review Data Tools Help	_ZUI1_Many.ctg] - [Linkage Configuration]						
Fixed Width	File 1:	N:\Test data	a 2011\KY201	11_test_Oct13.txt			
Data Type: Delimited	File 2:	N:\Medicaid	d\Enrollment	Data\Medicaid_Finder_2011.c			
Select Blocking Variables					Se	elect ID Variables (File 1)	
Data Item (File 1)	Data Iter	n (File 2)		Phonetic System		Study ID	
First Name	First_Na	me		NYSIIS			
Last Name	Last_Na	me		NYSIIS			
Social Security Name	SSN						
D	505						
Select Matching Variables and Me	thods				Se	elect ID Variables (File 2)	
Data Item (File 1)	Data Iter	n (File 2)		Matching Method		Medicaid_ID	
First Name	First Name First_Name		First Name		*		_
Last Name	Last_Na	me		Last Name			
* Social Security Name	SSN			SSN			
Middle Name	M			Middle Name			
Date of Birth	DOB			Date			
Gender	Gender			Exact			_
						Advanced.	
Missing Value (File 1)	Missing Value (File	2)	☑ Direct N	Method Best Match		Save	
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			C·\ReaPl	lus\LinkPlus\Report		Run	
Add Remove	Add	Remove	□ Genera	ate Non-match Report			
To learn how to select ID variables	s, please click on the	help icon in t	the tool bar				





Link Plus – Screen Shot

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7	1	1	106	46		00045	MOOTRY	DELMA	R	801426167 07201931	2					1502 HALGER DRIVE	KIRTLAND	RI
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7	1	2	105	86		00085	MAHAFFEY	AIMEE	P	803328262 06031922	2					508 1/2 PICKNEY CT	PAGE	RI
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0	4	25	82	48002	10048002	00023	PARRAN	MARGAR	ET E	805628903 10111935	2	906 2ND ST N # 7	FARMINGTON	(DE	99687	10100 300111 EAST	CIDADICON	IVIN
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Deterministic Data Linkage

- Following variables are included in the deterministic data linkage: SSN, date of birth, gender, last name, first name (both truncated to the first 6 letters).*
- Following process is used to identify matches:
 - Step 1: SSN, Last Name, First Name (Type 1)
 - Step 2: SSN, Last Name, Date of Birth (Month), Gender (Type 2)
 - Step 3: SSN, First Name, Date of Birth (Month), Gender (Type 3)
 - Step 4: First Name, Last Name, Date of Birth (Month and Year), Gender (type
 4)

*SM Koroukian. Linking the Ohio Cancer Incidence Surveillance System with Medicare, Medicaid, and Clinical Data from Home Health Care and Long Term Care Assessment Instruments: Paving the Way for New Research Endeavors in Geriatric Oncology. J Registry Manag. 2008 Winter; 35(4): 156–165.





Identifying True Matches

Medicaid linkage

- Combine "true" matches resulted from manual review process: one-to-many matches and many-to many.
- > Further check matches from the deterministic approach

Humana linkage

- ➤ Identify "true" matches resulted from manual review process (manyto many matches) by state
- > Further check matches from the deterministic approach
- Combine "true" matches from all states



Results – Medicaid (Probabilistic)

There are total 2644 true matches. One-to-many and Many-to-Many only add one extra true case individually.

		1 to M		M to M					
		N	% of		N	% of			
	N	(True	True	N	(True	True			
Score	(Matches)	Matches)	Matches	(Matches)	Matches)	Matches			
20+	2591	2591	100.0%	2591	2591	100.0%			
18-19.9	19	19	100.0%	19	19	100.0%			
15-17.9	23	23	100.0%	23	23	100.0%			
13-14.9	18	6	33.3%	18	7	38.9%			
11-12.9	123	4	3.3%	119	3	2.5%			
10-10.9	180	0	0.0%	175	0	0.0%			





Results – Medicaid (Deterministic)

➤ Total 2650 matches identified by deterministic matching with 2631 true matches. Sensitivity = 0.995, specificity = 0.998, PPV = 0.980, NPV=0.997

		N	% of True
Match Method	N (Matches)	(True Matches)	Matches
SSN, Last Name, First Name	2485	2485	100.0%
SSN, Last Name, Month of Birth, Gender	71	71	100.0%
SSN, First Name, Month of Birth, Gender	40	40	100.0%
Last Name, First Name, Month of Birth, Year of Birth, Gender	54	35	64.8%
Total	2650	2631	99.3%





Results – KY Humana (Probabilistic)

There are total 3825 true matches. 3784 from KY claims, 12 IN, 13 OH and 16 neighboring states. One-to-many from KY claims adds 5 extra true matches, and Many-to-Many from KY claims adds 17 extra true case separately.

	N	1 to M N (True	% of True	N	M to M N (True	% of True
Score	(Matches)	Matches)	Matches	(Matches)	Matches)	Matches
20+	3416	3414	99.9%	3416	3416	100.00%
18-19.9	93	90	96.8%	93	93	100.00%
15-17.9	176	159	90.3%	176	165	93.8%
13-14.9	211	88	41.7%	204	85	41.7%
11-12.9	694	25	3.6%	686	27	4.0%
10-10.9	978	4	0.4%	979	4	0.4%
9-9.9	2193	2	0.0%	2215	4	0.0%



Results – Other States Humana (Probabilistic)

		IN			ОН		Neighboring States			
Score	N Matches	N True Matches	% of True Matches	N Matches	N True Matches	% of True Matches	N Matches	N True Matches	% of True s Matches	
20+	13	13	100.0%	12	12	100.0%	16	16	100.0%	
18-19.9	3	1	33.3%	0	0	0.0%	0	0	0.0%	
15-17.9	7	0	0.0%	10	1	10.0%	10	0	0.0%	
13-14.9	43	0	0.0%	73	0	0.0%	73	0	0.0%	
11-12.9	323	0	0.0%	473	0	0.0%	473	0	0.0%	
10-10.9	642	0	0.0%	718	0	0.0%	718	0	0.0%	
9-9.9	1598	0	0.0%	1648	0	0.0%	1648	0	0.0%	





Results – Humana (Deterministic)

➤ Deterministic matching identified 3761 true matches out of total 3825. Excluding type4 matches from out of state claims, Sensitivity = 0.994, specificity = 0.990, PPV = 0.981, NPV=0.990

Match Method	Kent	uckv	Ir	ndiana		Ohio	Neighboring States	
	N	N_True		N_True	N	N_True	N	N_True
SSN, Last Name, First Name	3262	3262	13	13	11	11	16	16
SSN, Last Name, Month of Birth,								
Gender	112	112	1	1	0	0	0	0
SSN, First Name, Month of Birth,								
Gender	41	41	0	0	1	1	0	0
Last Name, First Name, Month of								
Birth, Year of Birth, Gender	418	346	50	0	98	0	273	0
Total	3833	3761	64	14	110	12	289	16



Summary

- Link Plus is relatively stable and fast.
- ➤ One to Many and Many to Many matching generated almost identical results. Many to Many may provide slight more matches.
- Cut-off values for linkage score can be set at 10 or higher. It is very rare a true cases can be identified with a value below 10. This depends on the quality of data, particularly the quality of SSN.
- ➤ Deterministic matching provides comparable result compared to the probabilistic data linkage.
- Choosing the best algorithm to use depends on many interacting factors, such as time, resources, data access, quality of matching variables.



If Manual Review is Not Possible

- "True" match if linkage score > 18 or it is a deterministic match
 - \triangleright Medicaid: sensitivity = 0.997, specificity = 0.998 PPV = 0.993, NPV = 0.999
 - \triangleright Humana: sensitivity = 0.999, specificity = 0.990 PPV = 0.982, NPV = 0.999
- ➤ This algorithm provides better linkage than the deterministic matching.



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