# A Probabilistic Geocoding System based on a National Address File

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Project web page: **http://datamining.anu.edu.au/linkage.html**

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## Outline

- Geocoding
- Geocoded National Address File (G-NAF)
- *Febrl* geocoding system
- Address cleaning and standardisation
- Processing G-NAF
- Geocode matching engine
- First results and geocoding examples
- Future work

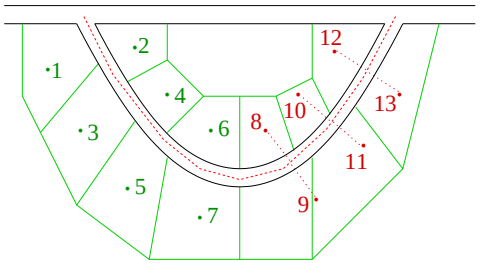


## Geocoding

- The process of assigning geographical coordinates (longitude and latitude) to addresses
- It is estimated that 80% to 90% of governmental and business data contain address information
  *US Federal Geographic Data Committee*
- Useful in many application areas
  - GIS, spatial data mining
  - Health, epidemiology
  - Business, census, taxation
- Various commercial systems available
  (e.g. MapInfo, www.geocode.com)



## Geocoding techniques



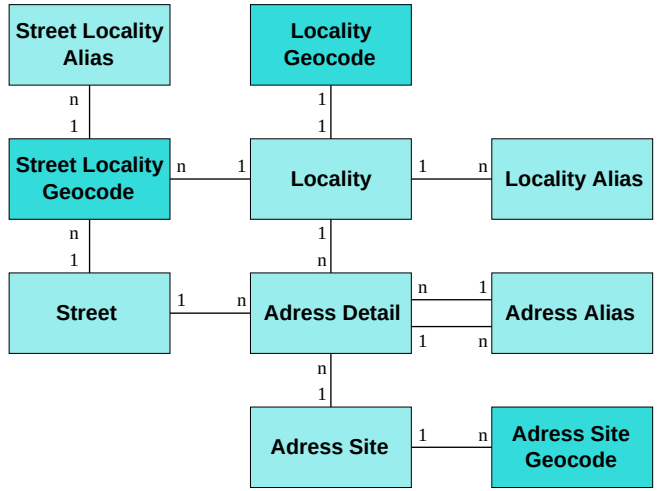

- Street centreline based (many commercial systems)
- Property parcel centre based (our approach)
- A recent study found substantial differences
  (specially in rural areas)
  *Cayo and Talbot; Int. Journal of Health Geographics, 2003*

## Geocoded National Address File

- Need for a national address file recognised in 1990
- 32 million source addresses from 13 organisations
- 5-phase cleaning and integration process
- Resulting database consists of 22 files or tables
- Hierarchical model (separate geocodes for each)
  - Address sites
  - Streets
  - Localities (towns and suburbs)
- Aliases and multiple locations possible



## Simplified G-NAF data model





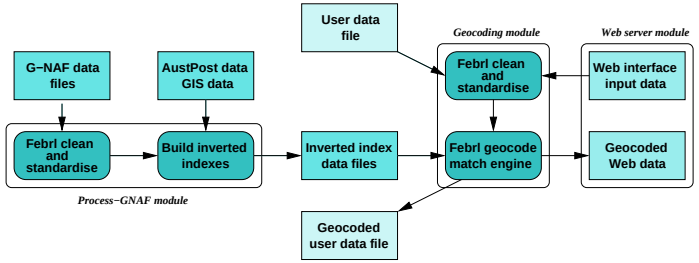

## G-NAF file characteristics

| G-NAF data file | Number of records / attributes |
|---|---|
| ADDRESS_ALIAS | 289,788 / 6 |
| ADDRESS_DETAIL | 4,145,365 / 28 |
| ADDRESS_SITE | 4,096,507 / 6 |
| ADDRESS_SITE_GEOCODE | 3,336,778 / 12 |
| LOCALITY | 5,017 / 7 |
| LOCALITY_ALIAS | 700 / 5 |
| LOCALITY_GEOCODE | 4,978 / 11 |
| STREET | 58,083 / 6 |
| STREET_LOCALITY_ALIAS | 5,584 / 6 |
| STREET_LOCALITY_GEOCODE | 128,609 / 13 |

- New South Wales data only



## Febrl geocoding system



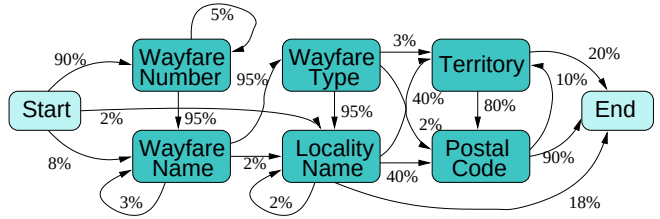

- *Febrl* (Freely extensible biomedical record linkage) (open source, object oriented, written in Python)
  - Experimental platform for rapid prototyping of new and improved linkage algorithms
  - Modules for data cleaning and standardisation, data linkage, deduplication, and geocoding

## Address cleaning and standardisation

- Real world data is often *dirty* (missing values, different coding formats, typographical errors, out-of-date data)
- For accurate geocode matching, we want clean data in well defined fields
- *Febrl* address cleaning is a three step process
  1. Input data is cleaned (make lower case, remove certain characters, correct misspellings and abbreviations)
  2. Split input into a list of words and numbers, then tag them (using rules and user definable look-up tables)
  3. Give tag lists to a probabilistic hidden Markov model (which assigns tags to output fields)



## HMM standardisation example




- Raw input: *'73 Miller St, NORTH SYDENY 2060'*
  Cleaned into: *'73 miller street north sydney 2060'*
- Word and tag lists:
  `['73', 'miller', 'street', 'north_sydney', '2060']`
  `['NU', 'UN', 'WT', 'LN', 'PC' ]`
- Example path through HMM
  `Start -> Wayfare Number (NU) -> Wayfare Name (UN) -> Wayfare Type (WT) -> Locality Name (LN) -> Postal Code (PC) -> End`



## Processing G-NAF

- Two step process
  1. Do cleaning and standardisation as discussed (to make G-NAF data similar to input data)
  2. Build inverted indices (sets, implemented as keyed hash tables with field values as keys)
     Example (postcode): `'2000':(60310919,61560124)`
- Within geocode matching engine, intersections are used to find matching records
- Inverted indices are built for 23 G-NAF fields



## Additional data files

- Use external *Australia Post* postcode and suburb look-up tables for correcting and imputing
  (e.g. if a suburb has a unique postcode this value can be imputed if missing, or corrected if wrong)
- Use boundary files for postcodes and suburbs to build *neighbouring region* lists
  - Idea: People often record neighbouring suburb or postcode if it has a higher perceived social status
  - Create lists for direct and indirect neighbours (neighbouring levels 1 and 2)

## Geocode matching engine

- Rules based approach for exact or approximate matching
- Start with address and street level matching set intersection
- Intersect with locality matching set (start with neighbouring level 0, if no match increase to 1, finally 2)
- Refine with postcode, unit, property matches
- Return best possible match coordinates
  - Exact / average address
  - Exact / many street
  - Exact / many locality / no match



## First results

| Match status | Number of records | Percentage |
|---|---|---|
| Exact address level match | 7,288 | 72.87 % |
| Average address level match | 213 | 2.13 % |
| Exact street level match | 1,290 | 12.90 % |
| Many street level match | 154 | 1.54 % |
| Exact locality level match | 917 | 9.17 % |
| Many locality level match | 135 | 1.35 % |
| No match | 3 | 0.03 % |

- 10,000 NSW *Land and Property Information* records
- Average 143 milliseconds for geocoding one record on a 480 MHz UltraSPARC II



## Geocoding examples



- Red dots: Febrl geocoding (G-NAF based)
- Blue dots: Street centreline based geocoding



## Future work

- Improve probabilistic data cleaning and standardisation
- Improve performance (scalability and parallelism)
- Improve matching algorithm
- Improve user interface (currently simple Web demo)
- Provide feedback on G-NAF to improve data quality
- Develop privacy preserving geocoding