

# Classifying consum ers: Adding more to the Census

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## Agenda



- Basic principles
- What were we thinking about?
- Why is interaction of variables critical?
- Measuring improvement
- Requests we turned down









## Basic principles



- People in sim illerareas have the same needs and lifestyles
- · You can find these people
- Without ever having metorheard of some body we aim to predict
  - Likely spending
  - Likely ownership of goods
  - Likely readership
  - Likely reactions to offers
- And get it substantially right









## Ourthinking



- . We to done this since the 70's
- One key change in the census changed ourthinking

OAsare more homogeneous







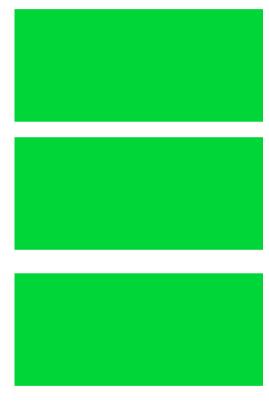


## Initial approach (1970's)



Initially data came in big blocks.

Census outputareas were pre-defined and you had to classify whatever mix ofpeople occurred.











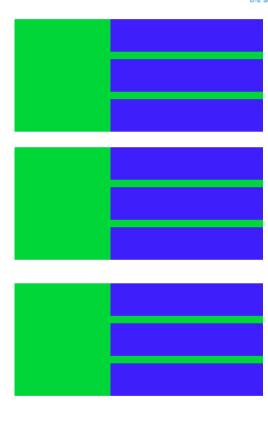
## Adding detail

acorn

Census areas could contain different people so postcode data.

Random noise and sampling issues were not considered serious. The "Shaw test"

Assum ed the balance of proof favoured the postcode











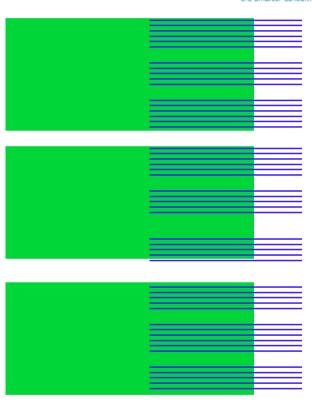
## Approach tums on its head!



Now the OA's are **designed** to be hom ogeneous.

So in theory all the postcodes are the same.

Balance of proof has shifted. It is no longer sufficient to be as casual about the incompleteness of postcode data.











## Mixing methods - two stages



## Stage 1

Classification of Census outputareas

## Stage 2

Introduce other data

Decision rules to re-classify
postcodes









## Advantages of two stages



- . Method albws easierannualupdates
  - Address new housing
  - Address infillhouse build
  - Address m igration
  - Focus on areas of change
- · Comparison with direct classification showed sharper targeting in all tests.









## Classification: How variables interact



- The information implicit in key variables makes others redundant
- Throwing too many variables into the pot makes classifications worse









## Types of variables



- Ranges
  - Household size
- Ratios
  - Roomsperhh
- Fractions
  - % dwellings with 1-2 rooms
- Incomplete sample
  - Bungabws

- Inverse variables
  - White
  - No religion
- Adapted ranges
  - Age quartiles



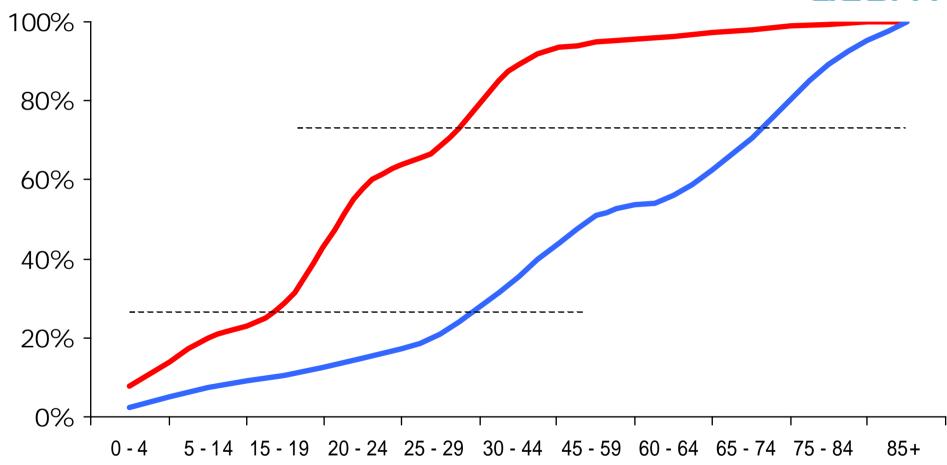






## Age quartiles













## A frequentm is understanding



- The type exists before any nam e or description
- Nam es can only provide a superficial picture
- Names may describe the majority behaviour, or highlight special features









## Specialcircum stances?



- If it is about an Asian type must Asian people live in the postcode?
- If it is labeled a student type must students live in the postcode?
- Is m anual intervention appropriate?
- W hich makes the classification better?







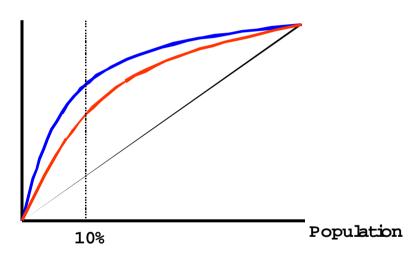


## The acid test: Measuring im provem ent



- Random lysplit consumer file
- Profile first.half
- Identify acorn types
- . Apply blind to other half
- Measure 10% pointofgains chart.
- Repeat frequently to assess standard deviation
- Measure statistical significance of in provement

#### Target Market











## Betterdiscrim ination



- Fam ily incom e
- . House price
- M ortgage
- Medicalinsurance
- Pensions
- Savings
- Investments
- Creditcards
- Newspapers
- . Media

- Cars
- . White goods
- Food spend
- Eating out
- W eekend breaks
- . Holidays
- . Music Opera/Art
- Com puters
- Internetusage









## Requests we turned down



• 'Sexy" names









### Nam es



- W hen asked over 95% of users opted for descriptive nam es
- Only advertising agencies were in favour of very evocative names
- Within our team we bved using funny names, but our MD banned us from taking about customers in this way."



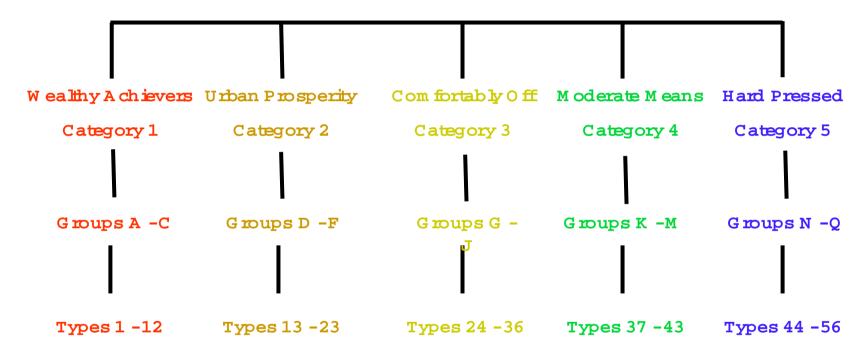






### acorn -structure













### 1: Affluent Achievers

# acorn

#### A W ealthy Executives

- A 1 W ealthy m ature professionals, large houses
- A.2 Wealthy working families with mortgages
- A 3 Villages with wealthy com muters
- A.4 Well-offm anagers, largerhouses

#### B AffluentGreys

- B 5 0 deraffluentprofessionals
- B.6 Farming communities
- B.70 bl people, detached hom es
- B.8 Mature couples, smaller detached homes

#### • C Flourishing Families

- C.9 Larger families, prosperous suburbs,
- C 10 Well-offworking families with mortgages
- C 11 W ell-offm anagers, detached houses
- C 12 Large fam ilies & houses in ruralareas









## A 1 W ealthy m ature professionals, large houses

- Typically found in Kingston, Stalbans, Guildford
- UnitTrusts
- Food spend
- Eam £50,000+
- High CreditCard spend
- Big houses
- Company Car
- FT
- Detached houses



## Requests we turned down



• 'Sexy" names

• 'Househod" acorn









## M ixing households and postcodes



- CACIhave had an individuallevel "acom" for over 4 years
- Branded as PeopleUK
- Marketing wanted to merge the brands
  - sam e nam es and types forboth
  - "its sounds such a great idea"
  - elegantsales and PR story
  - coordinate support material









## Well-offprofessionals, larger houses and converted flats



- OurChiefExecutive
  - Wealthy, lives in a big house, is a seniorm anagerand com mutes by car
  - Like a type 1, W ealthy m ature professionals, large houses
- A nearby neighbour
  - Affluent, rents a big flat, is a seniorm anager, and comm utes by tube
  - Like type 15, Affluenturban professionals, flats
- This type is in portant in a postcode classification
- A postcode can be a m ix
- A household can only be one thing or the other, it could never be this type









### Sum m ary



- C lassification is a m ix of skill and art
- Census changes prompted new thinking about our classification
- W e supplem ent the census with additional data
- Decision rules check the accuracy of each postcode classification
- This facilitates updates







