Exploring indicators of greenspace 'quality' in relation to self-reported general health

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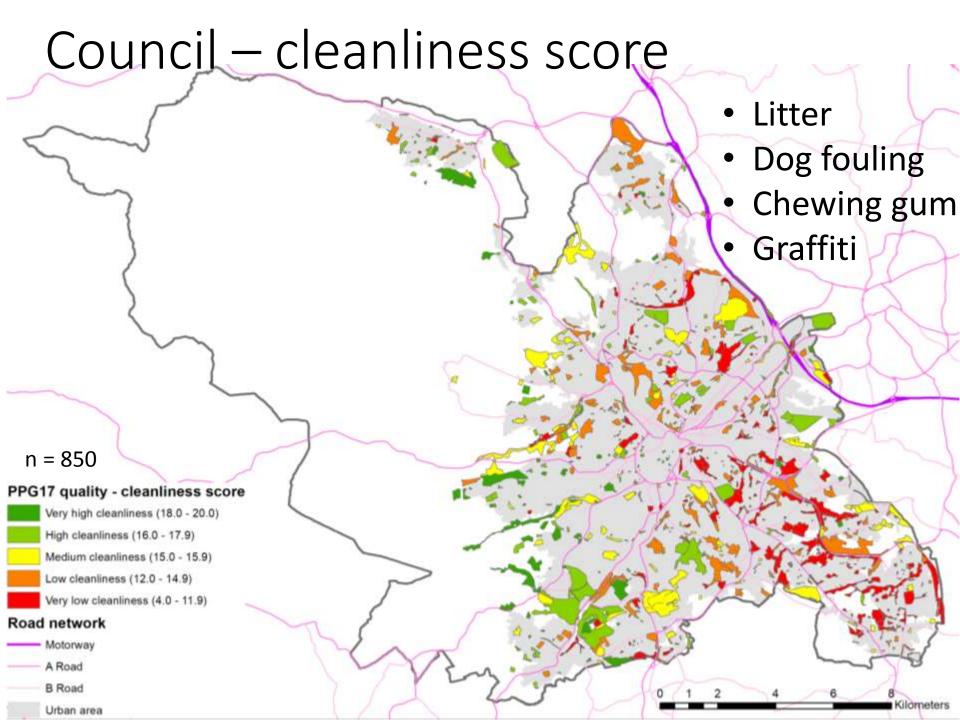
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Alternative measures of quality



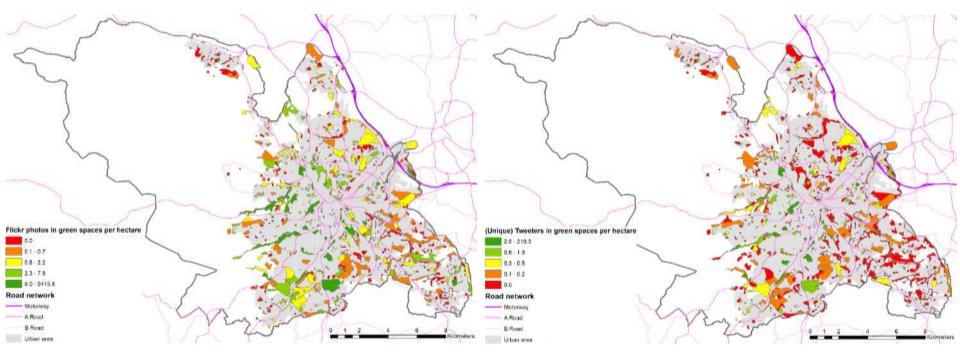
Flickr photos in greenspaces: 29,660 Average: 31 photos in each greenspace



1 yrs Tweets in greenspaces: 4,632

(Unique) Tweeters: 2,165

Average: 2.3 tweeters in each greenspace



spring winter leaves plants leaf
designed december yellow tree
fountain species using beauty crocus bird
flower sky flora squirrels white situated hot
wildlife bloom plant music nuts sunshine rose
pink day snow greysquirrel just July blossom

sheffield grey Art more robin purple britain portrait black

trees red grass today one nut statue Bear
land opened sun film blue bee most people acres
took pit gb notable water squirrel green
southyorkshire some
light pigeon topupthetea unitedkingdon
nature beautiful park flowers
unitedkingdom yorkshire england
november autumn time

southyorkshire yorkshire

Recreation england
clouds around Saturday through sledging
jump fence

Ground van taken
sunrise june back
sunsets over

Looking clones
bit Just Valley
some sky winter
city
see car
sunsets over
work

park

yalk

yalk

great summer

yiew
nice

try
home
sun playground
bike
grass

january

motorbike
fog pictures

children feb

abandoned music insect

sun path evidence crime background
flowers spring detritus wood before car rubbish Church
festival summer traces day started burgalry england
yellow csi go morning
snow some
building autumn tower April
uk yorkshire Dog bush Street outdoors green love full
gb Hat criminals flower Crowds early nature
purse significance police people burglary
Small

unitedkingdom year
afternoon scenery summer spring
highlandfling one SNOW frozen grass water
Pigs colour cliffhanger animals
transport Fling sycamore Highland film districts
animal farm animalfarm frost breeds lake day
rain outdoor over squirrel action festival square
sunset woodland landscape today color ducks
sun duck night eye using

UK old pond trees
light cold rare cow ice red building area saw public bird england sheep Donkey
South photo parkland nature cattle
october parkrun white protect

sheffield

Flickr sentiment analysis

- VADER (Valence Aware Dictionary and sEntiment Reasoner) Sentiment Analysis (version 2.5)
- a lexicon and rule-based sentiment analysis tool that is specifically attuned to sentiments expressed in social media

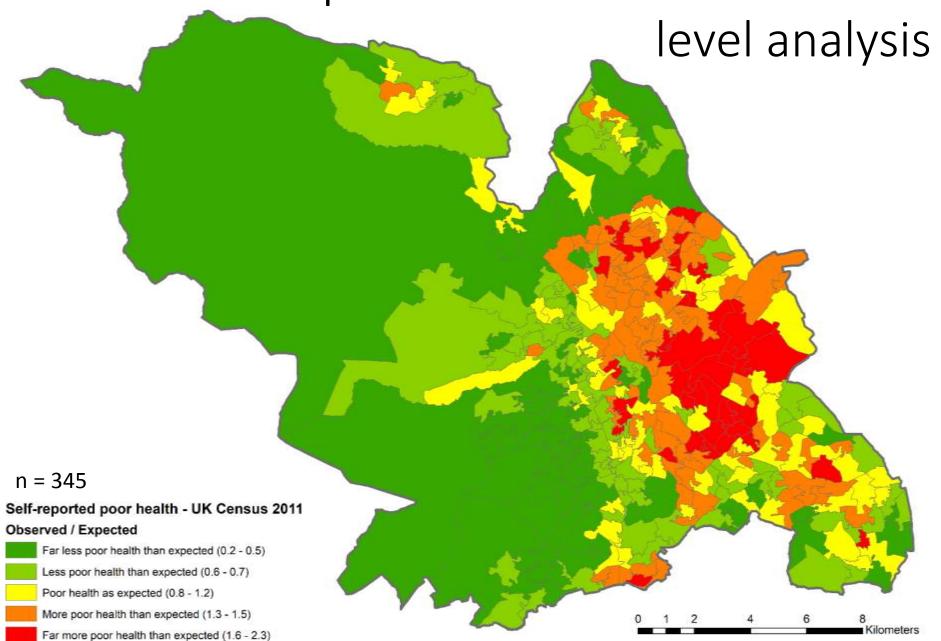
				was I was
Sentence:	negative	neutral	positive	
i love nature	0.00	0.19	0.81	
i don't love nature	0.63	0.37	0.00	
i love nature!	0.00	0.18	0.82	
i LOVE nature!	0.00	0.16	0.84	Flickr sentiment (positivernegative ratio) Very positive sentiment (4.0 - 140.5)
I love nature! :)	0.00	0.12	0.88	Positive sentiment (1.1 - 2.9) Medium sentiment (1.0) Negative sentiment (0.6 - 0.9) Very negative sentiment (0.0 - 0.5)
				Road network Motorway A Road B Road Uffan area 0 1 2 4 5 Schryele

Hutto, C.J. and Gilbert, E.E. (2014). VADER: A Parsimonious Rule-based Model for Sentiment Analysis of Social Media Text. Eighth International Conference on Weblogs and Social Media (ICWSM-14). Ann Arbor, MI, June 2014.

1. Greenspace level analysis

n = 850	Council - cleanliness score	Total number of Flickr photos	Flickr photos per hectare	Flickr sentiment (∑pos / ∑neg)	Total (unique) twitter users	Twitter users per hectare
Council - cleanliness score		0.06	0.08	0.03	0.08	0.08
Total number of Flickr photos	3 500		0.96	0.45	0.51	0.47
Flickr photos per hectare	- management			0.39	0.44	0.42
Flickr sentiment (∑pos / ∑neg)	4				0.32	0.29
Total (unique) twitter users	14	100				0.99
Twitter users per hectare						

2. Relationship with health: Census area



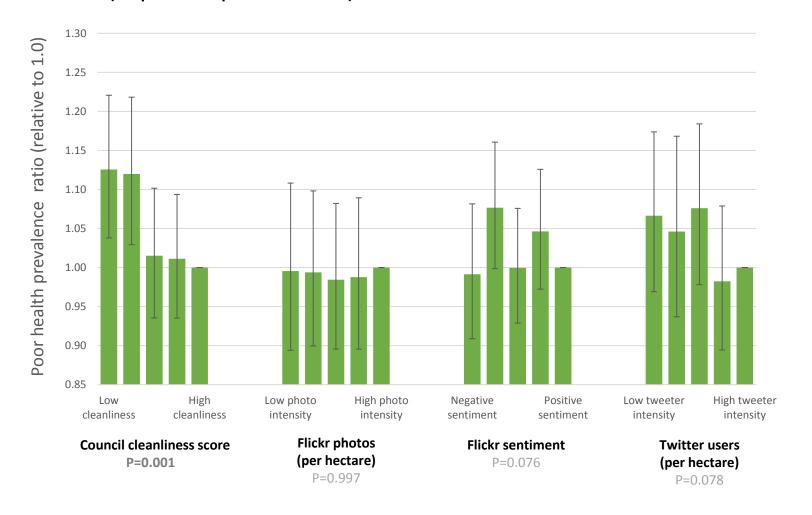
Negative binomial regression:

Dependent variable: Standardised poor health

Offset: In(expected poor health)

Confounders:

- Income deprivation
- Antisocial crimes
- Total green space
- Average garden size



Does quality matter?

Confounders:

Variable:	Quintile:	PR	95% CI
Income deprivation	1 (most deprived)	2.86	2.54, 3.21
	2	2.34	2.09, 2.62
	3	1.64	1.49, 1.82
	4	1.23	1.12, 1.34
	5 (least deprived)	1	
Antisocial crimes	1 (most crime)	1.05	0.94, 1.17
	2	1.06	0.96, 1.18
	3	1.07	0.97, 1.18
	4	1.10	1.01, 1.20
	5 (least crime)	1	
Total greenspace	1 (least green space)	1.09	0.99, 1.20
	2	1.04	0.97, 1.12
	3	1.02	0.95, 1.09
	4	1.00	0.93, 1.06
	5 (most green space)	1	
Average garden size	1 (smallest gardens)	1.22	1.11, 1.34
	2	1.11	1.02, 1.21
	3	1.08	1.00, 1.17
	4	1.08	1.00, 1.16
	5 (largest gardens)	1	

Quality measures:

Variable:	Quintile:	PR	95% CI
Council cleanliness	1 (least clean)	1.13	1.04, 1.22
score	2	1.12	1.03, 1.22
	3	1.02	0.94, 1.10
	4	1.01	0.94, 1.09
	5 (most clean)	1	
Flickr photos per	1 (low number of photos)	1.00	0.89, 1.11
hectare			
	2	0.99	0.90, 1.10
	3	0.98	0.90, 1.08
	4	0.99	0.90, 1.09
	5 (high number of photos)	1	
Flickr sentiment	1 (negative sentiment)	0.99	0.91, 1.08
	2	1.08	1.00, 1.16
	3	1.00	0.93, 1.08
	4	1.05	0.97, 1.13
	5 (positive sentiment)	1	•
Number of twitter	1 (low number of tweeters)	1.07	0.97, 1.17
users per hectare	_		
	2	1.05	0.94, 1.17
	3	1.08	0.98, 1.18
	4	0.98	0.89, 1.08
	5 (high number of tweeters)	1	

Conclusions

 Association found between greenspace cleanliness and self-reported poor health

 Social media derived measures of 'quality' were not similar to survey data derived from Sheffield City Council

 'Quality' measures derived from social media were poorly related to self-reported poor health

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