

## **The Clinical Effectiveness Group**

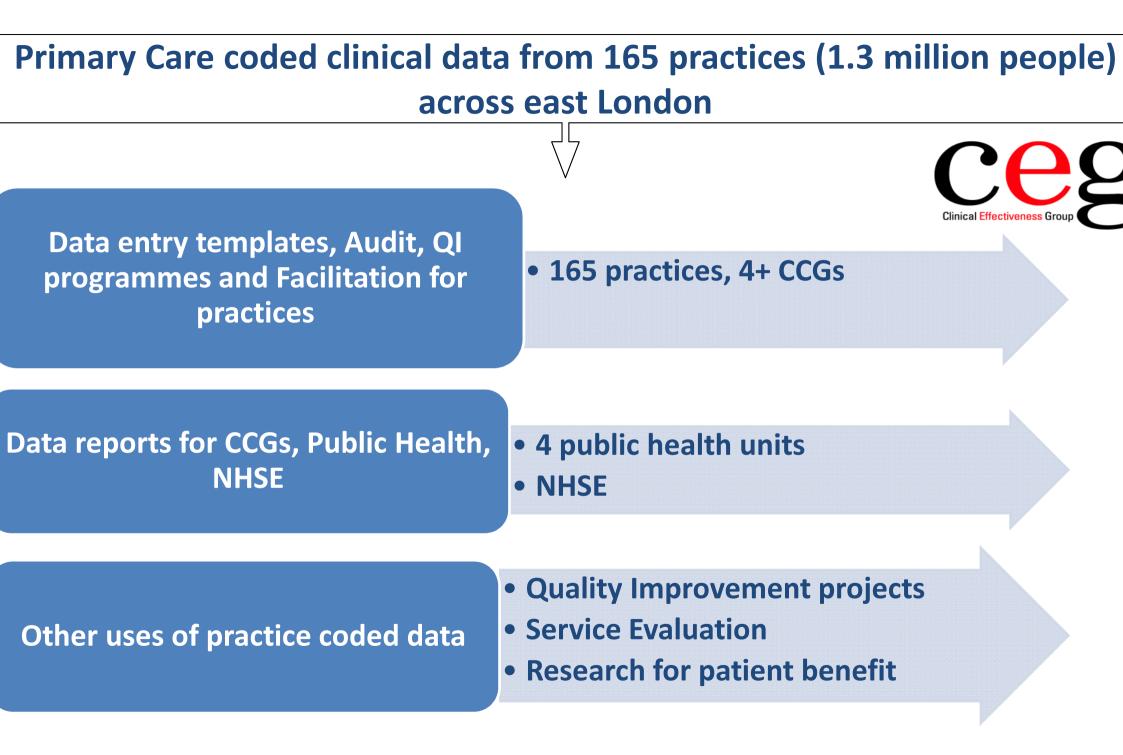
"A quality improvement organisation led by clinicians to support primary care in east London."



https://www.qmul.ac.uk/blizard/ceg/

Barts and The London School of Medicine and Dentistry

www.smd.qmul.ac

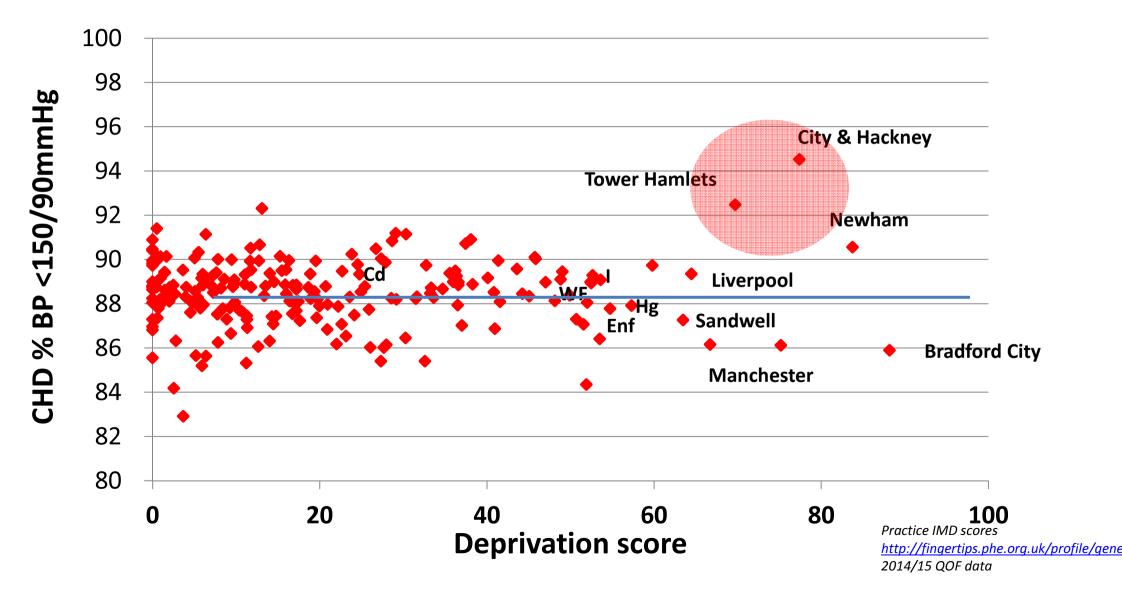


## Examples of working with GP practices to improve Quality of Care

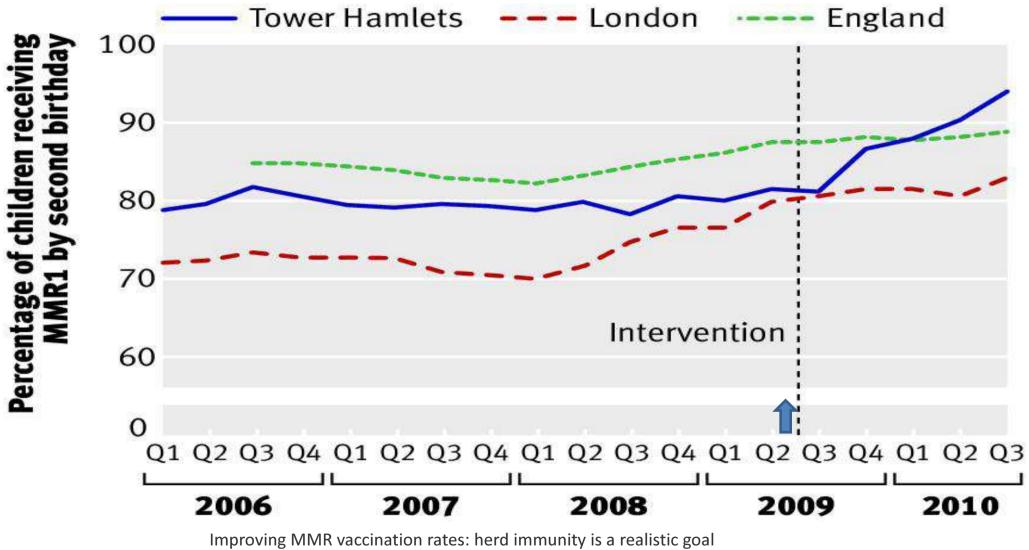
- 1. Using data to improve practice performance.
- 2. Practice networks: achieving improved clinical quality across practice clusters.
- 3. Working across primary and secondary care.



# Improving Blood Pressure managementEast London practice are the best in England



### MMR1 Immunisation 2006-10



Cockman P, Dawson L, Mathur R, Hull S, BMJ2011;343doi: 10.1136/bmj.d5703

## The NHS Long Term Plan January 2019

# Digitally-enabled primary and outpatient care will go mainstreater of the NHS

### CASE STUDY:

### Tower Hamlets Chronic Kidney Disease e-Clinics

Tower Hamlets CCG, working with City and Hackney and Newham CCGs, established e-Clinics to improve management of Chronic Kidney Disease and reduce End Stage Renal Disease. The new service supports timely provision of advice from the hospital specialist to the GP, to enable better management of the patient either in the community or with more specialist care where needed. A single pathway from primary to secondary care with rapid access to specialist advice provided by consultant led e-clinics have transformed the way the outpatient service is delivered. Since the e-Clinic began in December 2015, 50% of referrals are managed without the need for a hospital appointment. The average waiting time for a renal clinic appointment has fallen to five days, from 64 days in 2015.



### spected COVID-19 in primary care: how GP cords contribute to understanding differences in evalence by ethnicity.

lly Hull, Crystal Williams, Chris Carvalho, Mark Ashworth, mbiz Boomla

ne 2020

Barts and The London

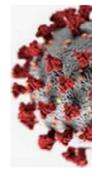
School of Medicine and Dentistry



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## **Study Objectives**

- To document the numbers and time-course of suspected COVID-19 presenting to east London general practice during the London epidemic up to the end of April 2020.
- To report disease prevalence by ethnic group, and explore whether observed differences can be explained by health data recorded in the GP record



## Methods

Cross sectional study, comparing characteristics of adults with and without COVID-19 codes in the GP record.

Based in 4 CCGs (165 practices) in east London, adult population of 1.2 Million

Data was extracted from GP computer systems after 30 April



## **Methods:** what data did we extract from GP records?

**Demographics:** Age in years, sex, ethnicity, social deprivation, CCG, practice identifier

#### **Co-morbidities:** 15 conditions in the Quality and Outcomes Framework

sthma, Chronic Obstructive Pulmonary Disease, Atrial Fibrillation, Cardio Vascular Disease, leart Failure, Hypertension, Peripheral Arterial Disease, Stroke/Transient Ischaemic Attack , liabetes, Chronic Kidney Disease, Depression, Severe mental illness, Epilepsy, Dementia, Cancer

## **Other Clinical Measures:** Body Mass Index, Current smoking status,



g. 1. Comparing test-positive cases in all of London, with those in the east London study area. ney occur at the same time, suggesting east London is representative of London as a whole.

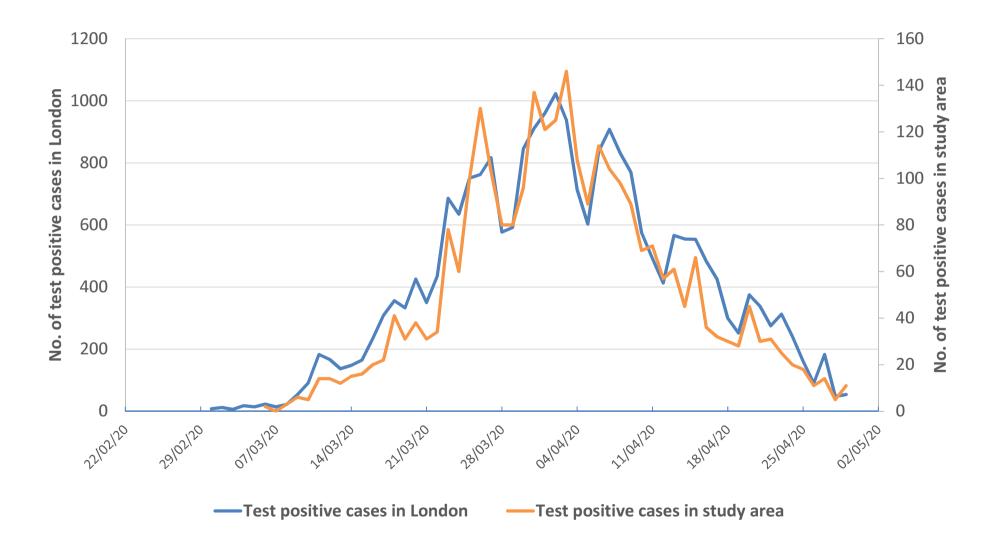
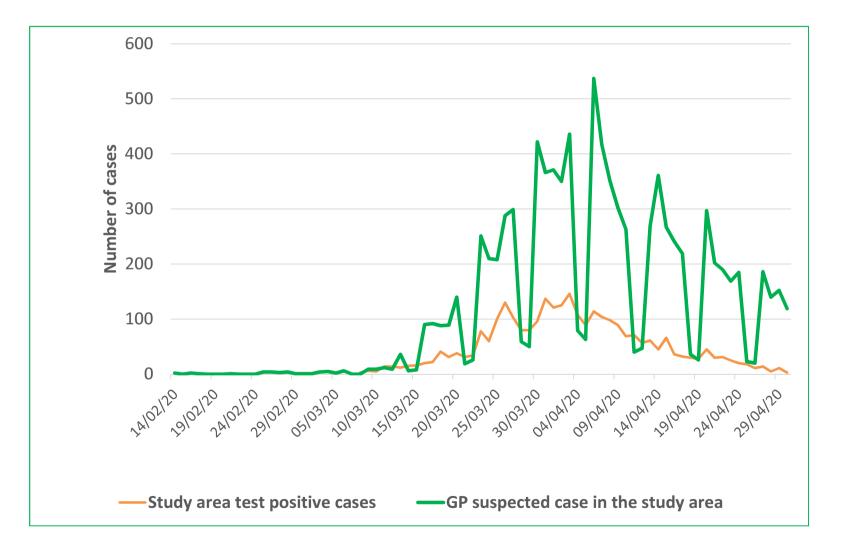
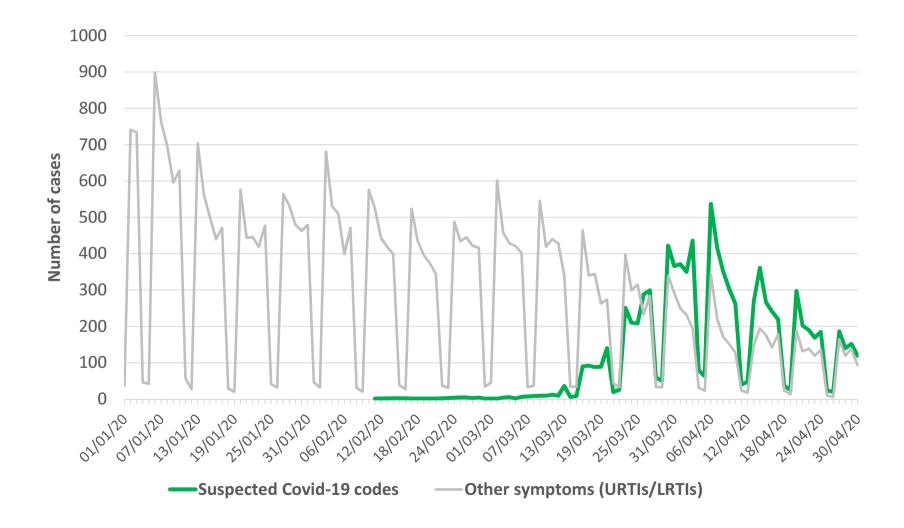


Figure 2. Comparing study area test-positive cases with GP coded suspected COVID cases These have a similar distribution,

This suggests that GPs have good discrimination for COVID presentations.



#### Figure 3. Comparison of GP suspected Covid-19 cases with GP recorded URTI/LRTI codes This shows the seasonal decline in URTI, compared to the rise of COVID, Suggests GPs are discriminating between URTI and COVID



#### Characteristics of those with and without suspected COVID-19 codes to the end of April 2020.

(Includes 1,257,136 patients aged ≥ 18 years from 165 practices)

	Suspected	Without	Univariate OR
	COVID-19	suspected	(95%CI)
	(%)	COVID-19 (%)	
tal	8,985	1,248,152	
e			
8-49 (ref)	5,134 (57.1)	926,886 (74.3)	
-69	2,723 (30.3)	235,616 (18.9)	2.18 (2.08-2.29)
er 70	1,128 (12.6)	85,650 (6.9)	2.45(2.29-2.62)
x			
ale (ref)	3,982 (44.3)	632,082 (50.6)	
male	5,003 (55.7)	616,070 (49.4)	1.28 (1.22-1.33)
hnicity			
hite (ref)	2,890 (32.2)	476,302 (38.2)	
uth Asian	2,859 (31.8)	259,464 (20.8)	1.98 (1.86-2.09)
ack	1,642 (18.3)	153,240 (12.3)	1.88 (1.77-2.00)
itional IMD 2015			
east deprived (ref)	30 (0.3)	8964 (0.7)	
	96 (1.1)	24029 (1.9)	1.35 (0.88-2.06)
	485 (5.4)	99395 (8.0)	1.22 (0.83-1.79)
	3557 (39.6)	541773 (43.4)	1.53 (1.05-2.23)
nost deprived	4807 (53.5)	560245 (44.9)	1.88 (1.29-2.74)

QOF long term conditions	Suspected COVID-19 (%)	Without suspected COVID-19 (%)	Univariate OR (95%Cl)
0 (ref)	3,740 (41.6)	881,460 (70.6)	
1	2,461 (27.4)	226,961 (18.2)	2.41 (2.29-2.54)
2	1,350 (15.0)	81,093 (6.6)	3.75 (3.52-3.99)
3	690 (7.7)	33,497 (2.7)	4.6 (4.25-5.02)
4+	744 (8.3)	25,141 (2.0)	6.5 (6.00-7.05)
Current smoker	1,047 (11.7)	217,396 (17.4)	0.60 (0.56-0.63)
Asthma	1,512 (16.8)	111,641 (8.9)	1.92 (1.81-2.03)
Atrial Fibrillation	248 (2.8)	10,299 (0.8)	3.16 (2.78-3.59)
Cancer	429 (4.8)	22,989 (1.8)	2.50 (2.26-2.75)
Coronary heart disease	504 (5.6)	23,114 (1.9)	2.98 (2.72 -3.26)
Chronic kidney disease (3-5)	716 (8.0)	32,203 (2.6)	3.11 (2.88-3.37)
COPD	331 (3.7)	14,467 (1.2)	2.92 (2.61-3.26)
Dementia	258 (2.9)	4,442 (0.36)	7.37 (6.48-8.39)
Depression	1,811 (20.2)	121,290 (9.7)	2.15 (2.04-2.27)
Diabetes	1,696 (18.9)	79,445 (6.4)	3.31 (3.13-3.49)
Epilepsy	157 (1.8)	10,321 (0.8)	2.00 (1.70 - 2.34)
Heart Failure	234 (0.7)	8,039 (0.6)	3.75 (3.28-4.28)
Hypertension	2,290 (25.5)	131,318 (10.5)	2.85 (2.71-2.99)
Learning disability	70 (0.8)	4,660 (0.4)	1.89 (1.49-2.40)
Severe Mental Illness	250 (2.8)	17,322 (1.4)	1.88 (1.65-2.13)
Peripheral arterial disease	87 (1.0)	3,608 (0.3)	3.00 (2.41-3.71)
Stroke TIA	284 (3.2)	11,514 (0.9)	3.24 (2.87-3.65)

#### ultivariate model for predictors of COVID-19 for adults aged $\geq$ 18 years (*n*=1,257,137 cases contributing to the m

		Model includes Demographic and Clinical factors		
		OR <sup>a</sup>	95% CI	P Value
	Male (ref)	1.00		
	female	1.17	(1.12 to 1.22)	<0.001
bands rs)	18-49 (ref)	1.00		
	50-69	1.30	(1.23 to 1.37)	<0.001
	>=69	1.25	(1.16 to 1.35)	<0.001
licity <sup>c</sup>	White (ref)	1.00		
	South Asian	1.93	(1.83 to 2.04)	<0.001
	Black	1.47	(1.38 to 1.57)	<0.001
rnal 2015 itiles <sup>c</sup>	1 least deprived (ref)	1.00		
	2	1.18	(1.09 to 1.28)	<0.001
	3	1.16	(1.07 to 1.25)	<0.001
	4	1.21	(1.17 to 1.37)	<0.001
	5 most deprived	1.26	(1.17 to 1.37)	<0.001

		Model (cont)		
		ORª	95% CI	P Val
Long term conditions	0 (ref)	1.00		
	1	1.77	(1.67 to 1.87)	<0.0
	2	2.28	(2.13 to 2.45)	<0.0
	3	2.60	(2.37 to 2.85)	<0.0
	>=4	3.67	(3.33 to 4.03)	<0.0
BMI, Kg/m²	Normal weight (ref)	1.00		
	Underweight	0.84	(0.73 to 0.97)	0.0
	Overweight	1.31	(1.24 to 1.38)	<0.0
	Obese	1.73	(1.63 to 1.84)	<0.0
	Morbidly Obese	2.20	(2.01 to 2.47)	<0.0

### Discussion

. The report demonstrates practice activity during the pandemic. The records of suspected cases can be used to detect new outbreaks on the future.

. Our data is at the early stage of disease trajectory. Ve see the same differences in risk of disease by ethnicity as the studies /hich look at death rates.

. Clinical factors (comorbidities) contribute only a small part of the ifference in risk of COVID by ethnicity. Further understanding needs ousehold composition, occupational and other risk factors.



## ower Hamlets General Practice OVID-19 Situation Report

ta obtained from Barts Health and GP practices through e Discovery Data Service st wave data till 2nd June 2020



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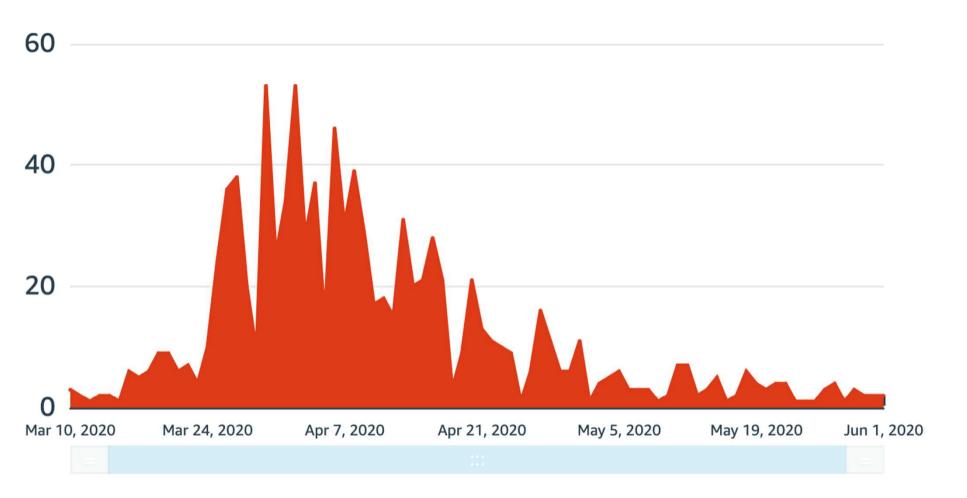
Barts and The London





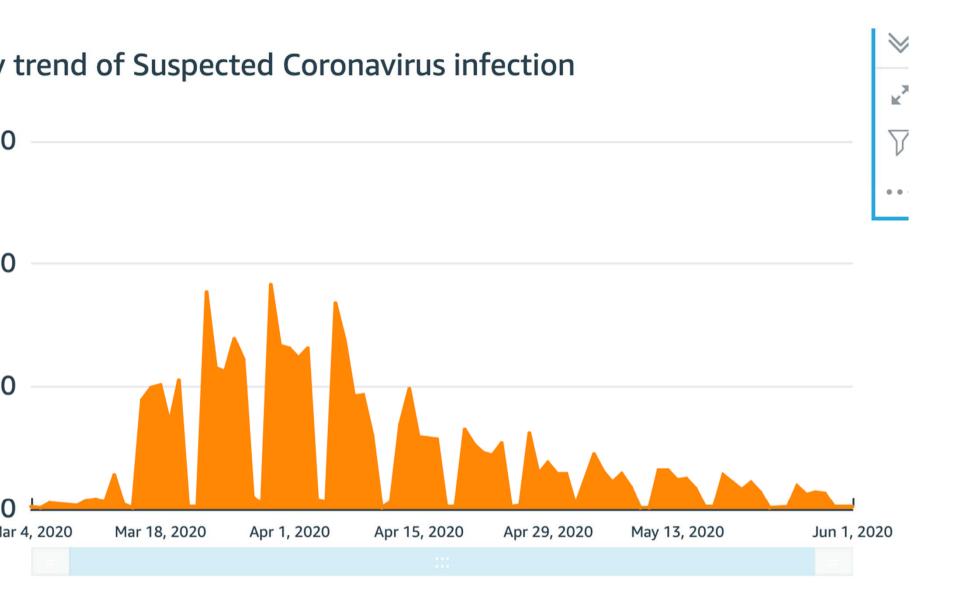
#### www.smd.qmul.ac

#### Day trend of Confirmed Covid 19



Confirmed Covid 19

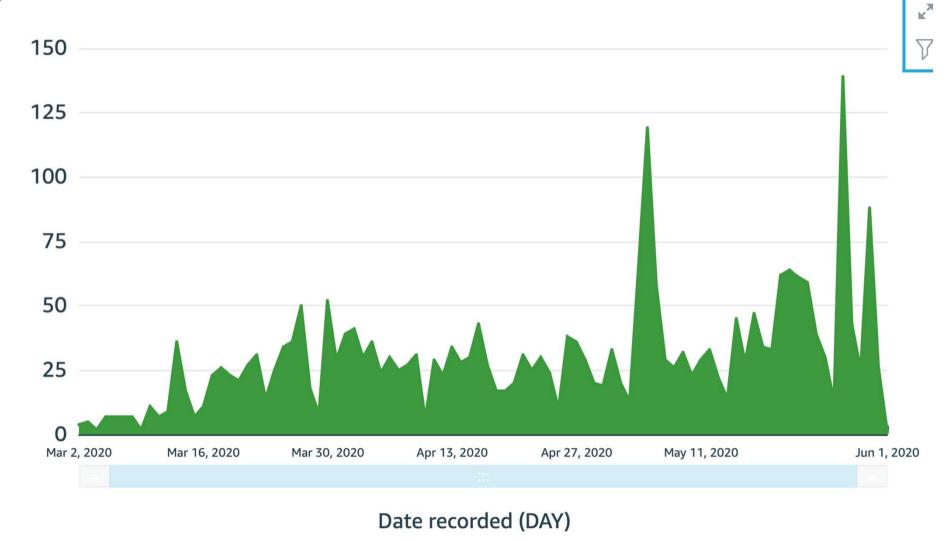




uspected coronavirus infection



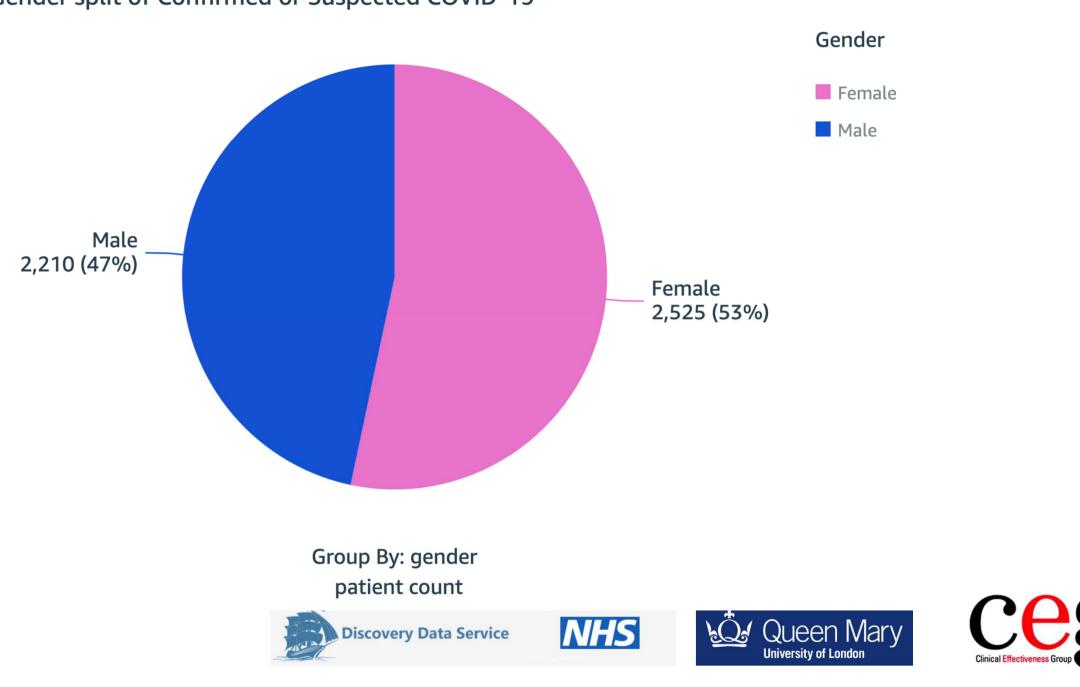
#### Day trend of Tested for Coronavirus infection



✓ Tested for coronavirus infection

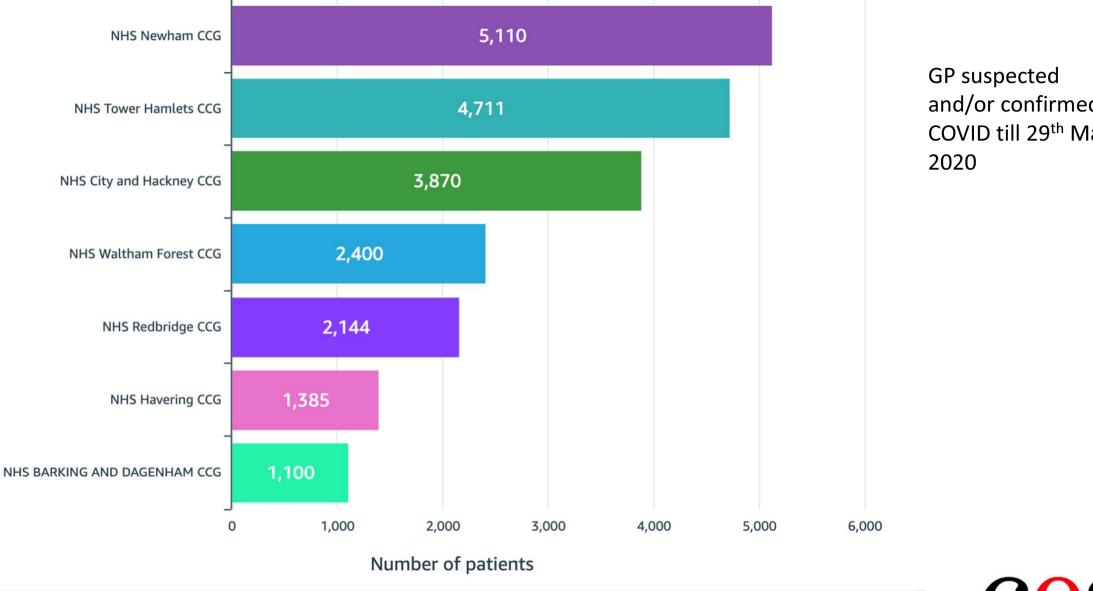


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#### Gender split of Confirmed or Suspected COVID-19

#### ount of patients by coronavirus category by CCG (use filters to drill down)



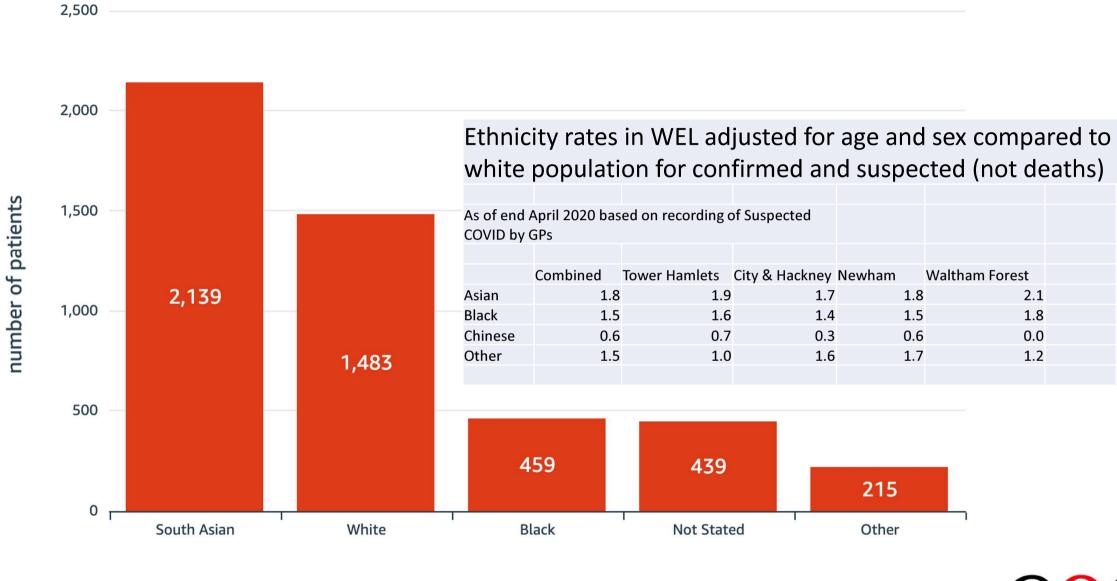








#### Ethnicity groups of Confirmed or Suspected COVID-19

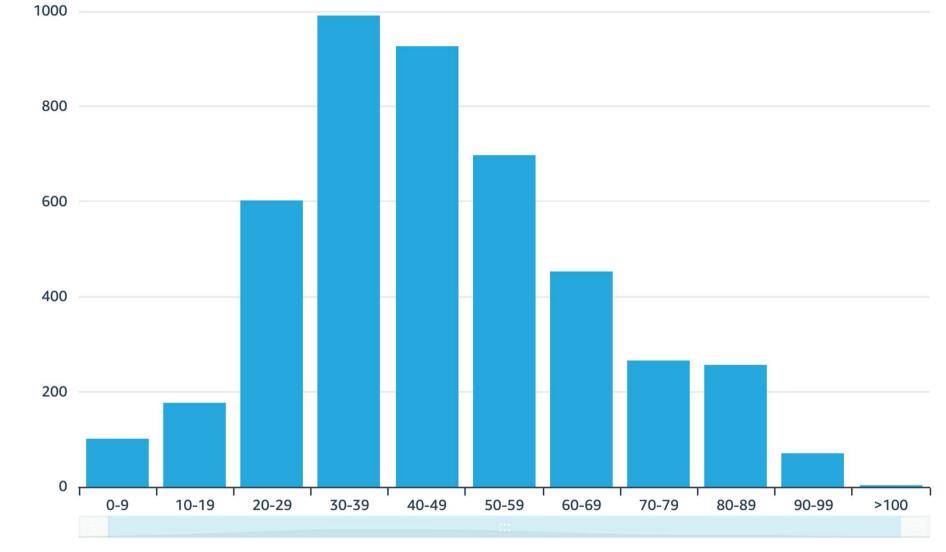








#### Age groups of Confirmed or Suspected COVID-19







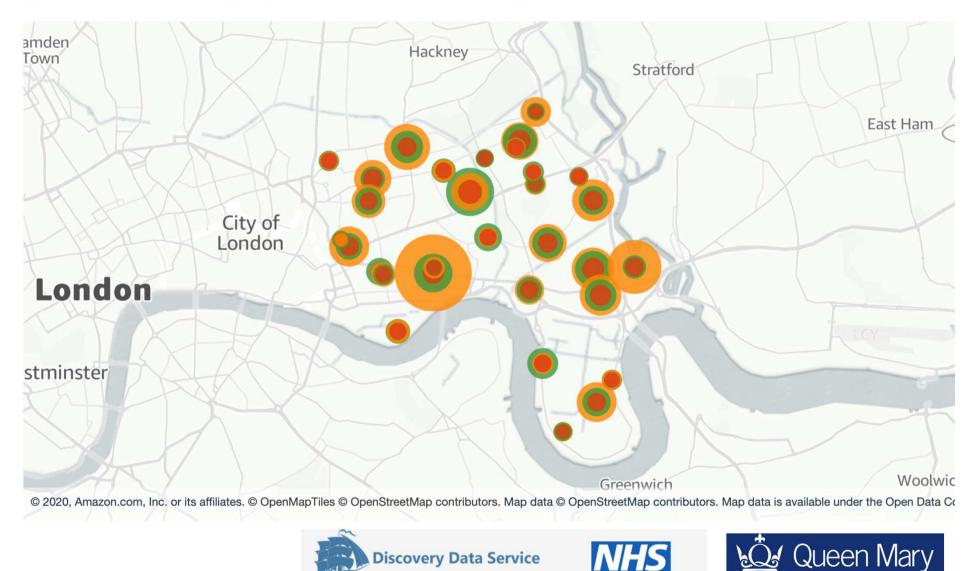
number of patients

#### Count of patients with Confirmed, Suspected and Tested for COVID-19

SHOWING TOP 32 IN LATITUDE. LONGITUDE AND TOP 3 IN CORONA STATUS

Confirmed Covid 19

Suspected coronavirus infection

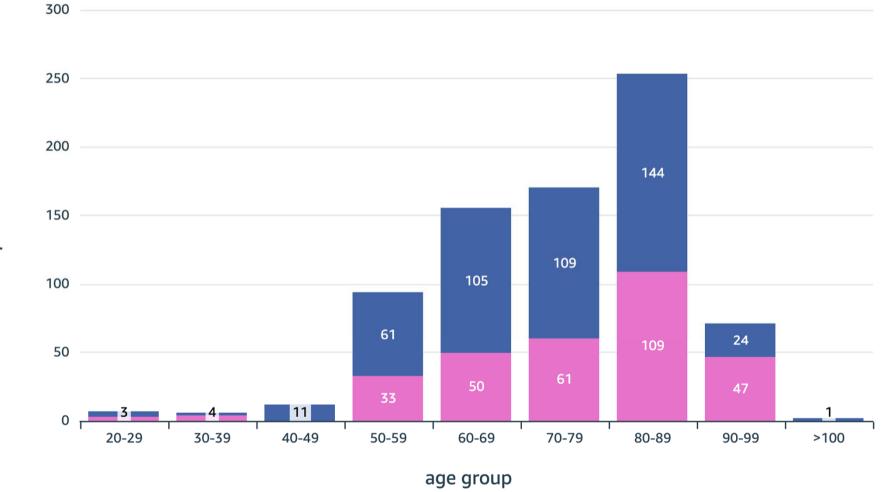


have been mapped to t GP practice postcodes. We will soor mapping to **Unique Prop** Reference Number (UP with blurring preserve confidentiali

These cases



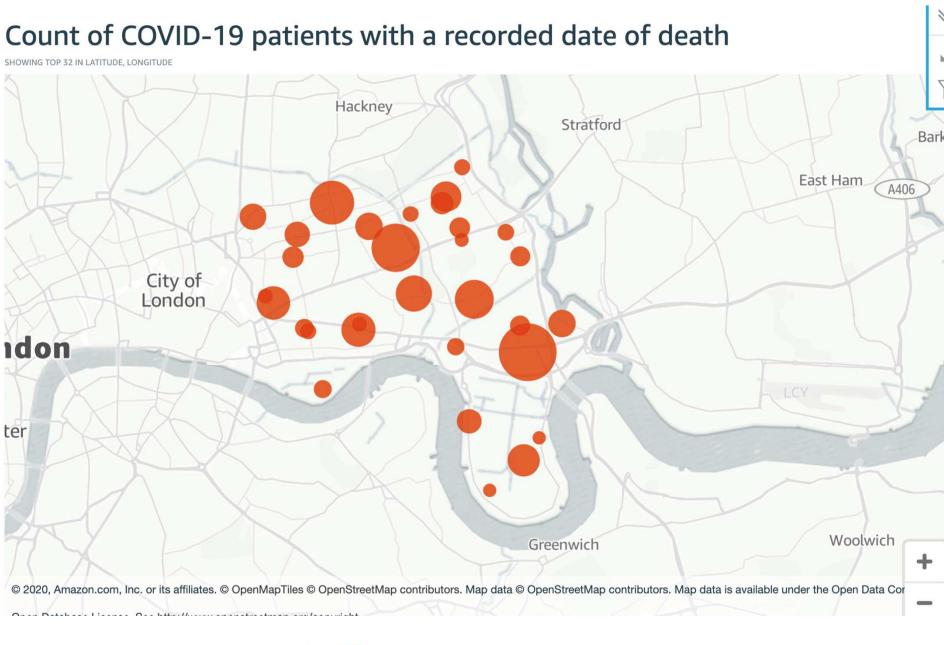
University of London



#### Count of deceased COVID-19 patients by age/sex (use filters to drill down)

Female Male





These cases have been mapped their GP pract postcodes. We will soon b mapping to Unique Prope Reference Number (UPR with blurring to preserve confidentiality

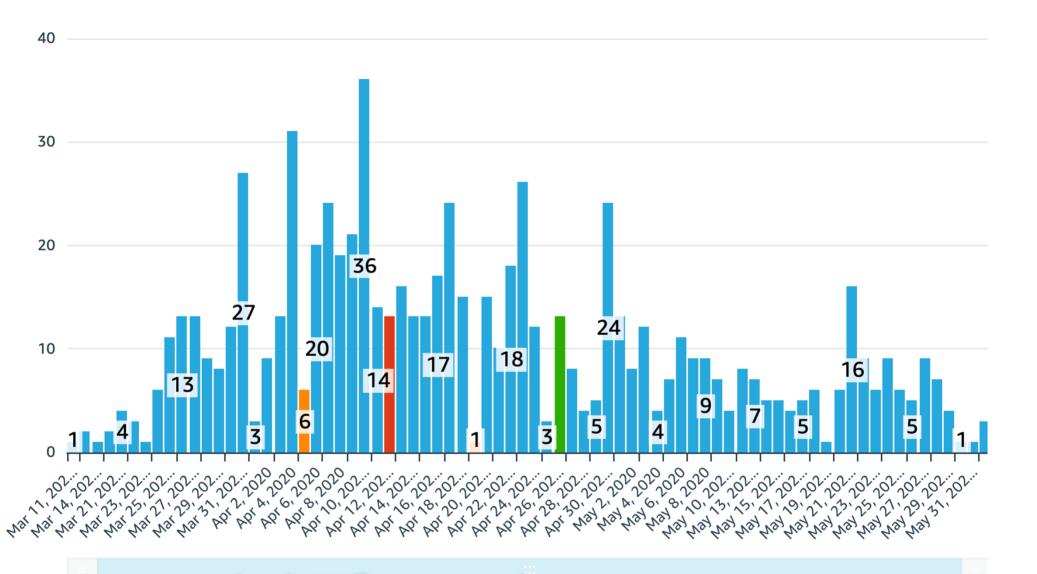








nt of deceased COVID-19 patients by day (use filters to drill down)





#### Count of deceased COVID-19 patients, by most common pre-existing condition

#### ethnic\_name

	White	South Asian	Black
Pre-existing condition	patient count $\stackrel{\text{In}}{\rightarrow}$	patient count	patient count
Essential hypertension	33	23	15
Chronic kidney disease stage 3	22	12	8
Type 2 diabetes mellitus	16	26	11
Chronic obstructive pulmonary disease	10	7	1
Atrial fibrillation	9	3	2
Heart failure	8	1	4
Asthma	7	7	2
Hypertensive disease	5	4	1
[V]Palliative care	5	4	3
Ischaemic heart disease	4	6	
Total	119	93	47



