

# PREDICTING AND CHARACTERISING USER IMPACT ON TWITTER

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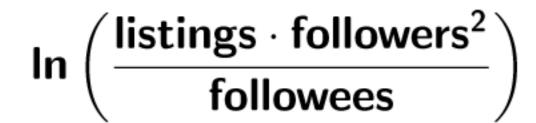
## IMPACT DEFINITION

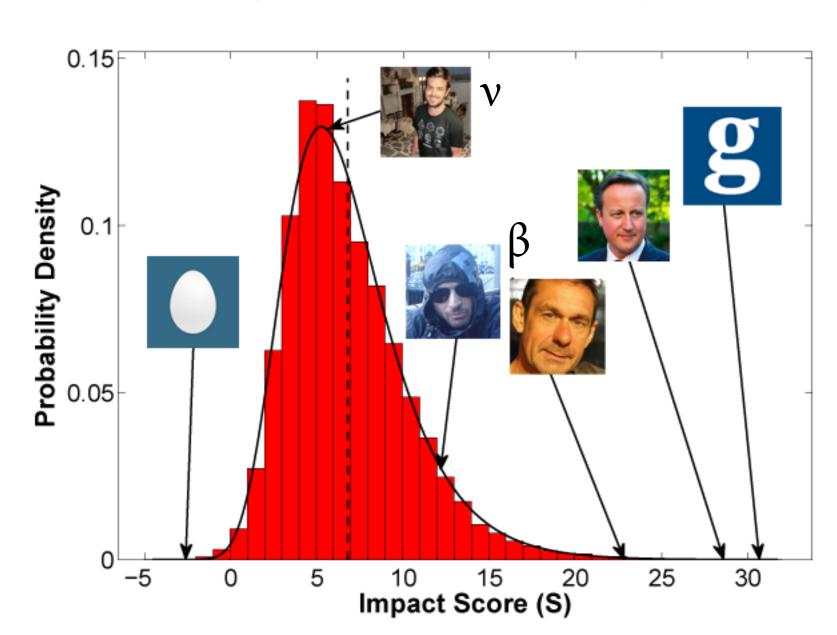
The

University

Sheffield.

ν,δ



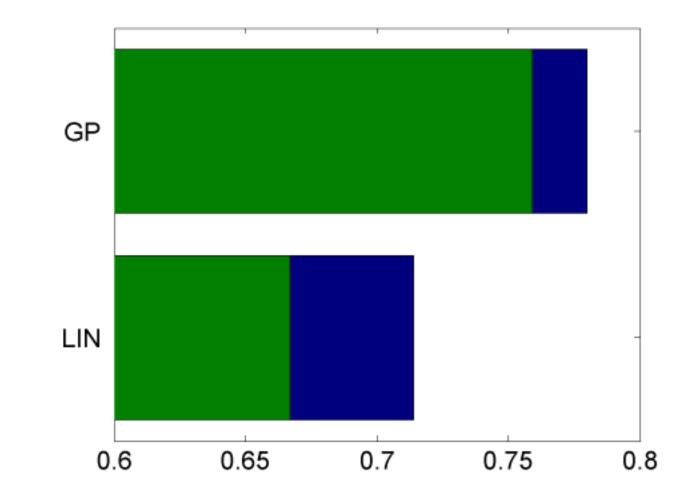


#### IMPACT PREDICTION

Task: Predict user impact based on features under the user's control

#### Ridge regression (LIN) Non-linear methods (GP)

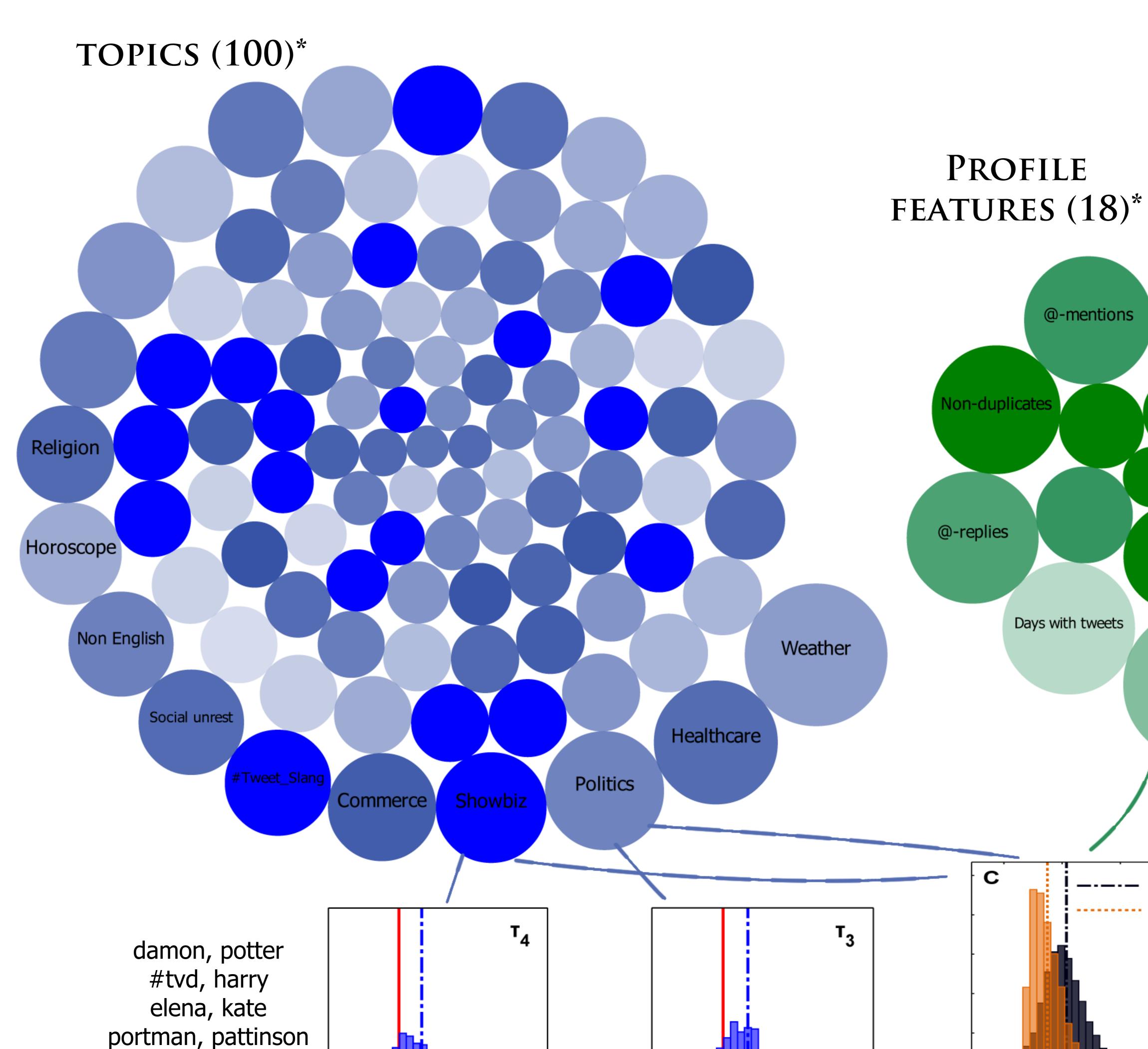
Gaussian Processes with ARD kernel



Green represents correlation using profile features, blue is the gain adding topics

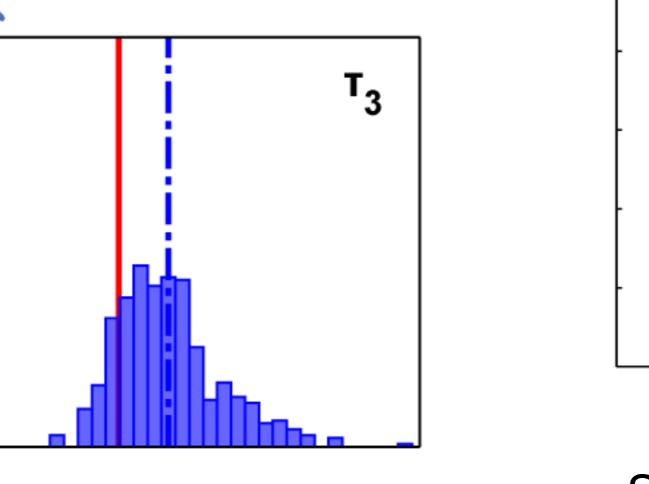
### DATA

38,020 UK located Twitter users from 14.04.2011-12.04.2012 ~48 mil deduplicated tweets

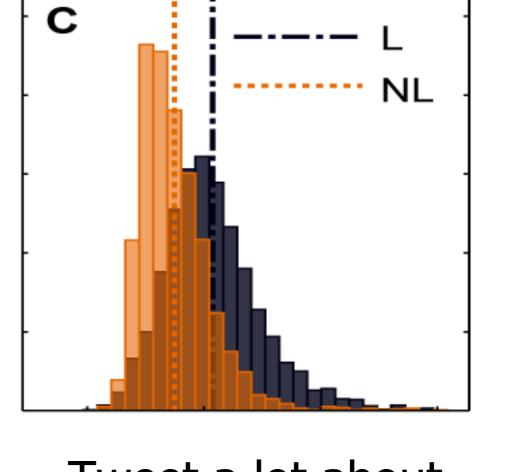


Topics computed over reference Twitter corpus Spectral clustering with NPMI as similarity measure

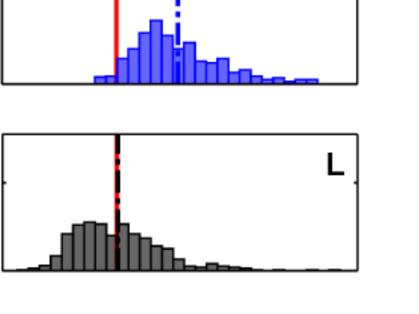
hermione, jennifer



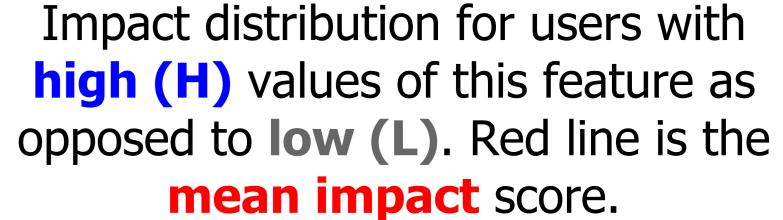
senate, republican gop, police arrested, voters robbery, democrats presidential, elections



Tweet a lot about Showbiz and Politics, with (L) or without (NL) using <URL>'s



tweets





Custom avatar

#hashtags

Population

<URL>

tweets\_ALL

Unique @-mentions

<sup>\*</sup> bubble size inverse proportional to learned GP ARD kernel lengthscales and represent predicitve relevance; colours split between the types of features; nuances of the same colour are only for visual effect