Figure 9: Number of bots detected by DeBot per day. Gaps indicate downtime due to update and maintenance.
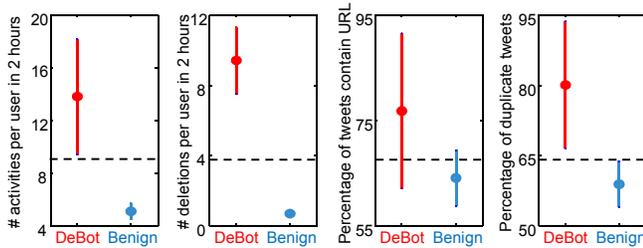

Figure 10: Comparison between benign accounts and accounts we (*DeBot*) detect as bot. The dashed lines show near complete separation in all of the features between benign accounts and the accounts *DeBot* detected considering the mean and the standard deviation.
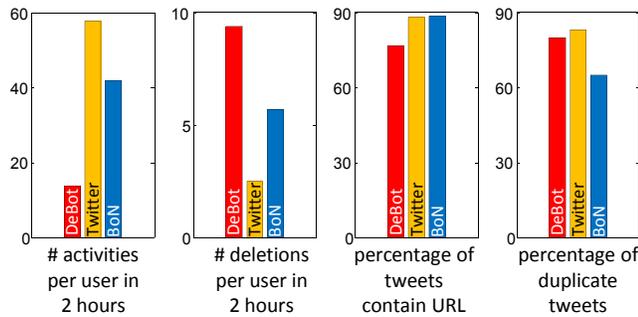

Figure 11: Comparison between accounts we detected as bot (*DeBot*), accounts suspended by Twitter and accounts detected as bot by Bot or Not? (*BoN*)

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