# Using Facial Recognition to gather Social Media Intelligence

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# 1 Literature Review

## 1.1 Background

### 1.1.1 SOCMINT

Social media intelligence (SOCMINT) is an emergent field in intelligence gathering where data is gathered from social media profiles. Massive amounts of data are added to social media services every day (Omand et al., 2012), much of it personal, making social media sites a potentially valuable resource when gathering information about groups or individuals (Ruiz, 2018). Social networks have also been used as a means of communication between persons of interest to the security services, making mining intelligence from their profiles a high priority (Omand et al., 2012)(Kilburn and Krieger, 2014).

After the 2011 riots in London that were organised in large part on social media, Her Majesty's Inspectorate of Constabulary stated that the police services were "insufficiently equipped" to effectively use SOCMINT in their response (Antonius and Rich, 2013) which suggests that social media intelligence sources may be woefully underutilised. This is not to say that the value of SOCMINT is not realised however, as many intelligence agencies are investing in tools to effectively gather and analyse SOCMINT (Antonius and Rich, 2013) or are performing case studies in to potential uses (Klontz and Jain, 2013).

While traditional human intelligence (HUMINT) focuses on building rapport and a foundation of trust in order to extract information from people of interest (Russano et al., 2014), users of social networking websites are much more likely to divulge personal information due to a misplaced senes of privacy (Livingstone, 2008). This makes SOCMINT attractive when attempting to gather data with little investment. The amount of data available to gather is vast in comparison to HUMINT sources (Omand et al., 2012), making mass collection and analysis viable (Unknown, 2013). The nature of SOCMINT makes it easier to analyse than HUMINT, which relies on "tells" and small social cues (Russano et al., 2014).

### 1.1.2 Uses of SOCMINT

As previously stated, SOCMINT has seen some emergent use particularly in the security services. The Greek Ministry of Defence has developed a framework to identify individuals fitting certain psychiatric profiles from their social media accounts to allow for early identification of potential insider threats (Kandias and Stavrou, 2015). By identifying factors that multiple intelligence agencies agree make a person more likely to pose an insider threat or negatively influence society (See appendix A), they were able to map usage habits (intensity, content, popularity) to these factors to draw conclusions about clusters of users. So far, the research has been helpful in insider threat prevention, delinquent behaviour prediction and forensic analysis support.

### 1.1.3 Facial Recognition

Facial recognition is a much more mature area of research than SOCMINT with many examples of industry usage. Facebook uses facial recognition to automate "tagging" photos with the identity of the persons pictured (Becker and Ortiz, 2008), and large companies are now releasing datasets such as YouTube Faces (Cui et al., 2013) in an effort to advance the field.

This is not to say that facial recognition is not without controversy however, as many privacy advocates have pointed out that accurate face regonition could infringe on their right to privacy (Ruiz, 2018). David Wood and Lucas Introne have posed that accurate facial recognition could lead to increased levels of surveillance, with no way to "opt out" (Introna and Wood, 2002)(Bowyer, 2004).

Maybe write about PRISM?

### 1.1.4 Uses of Facial Recognition

Facial recognition has many practical applications that are already being realised. As noted previously, Facebook uses facial recognition when "tagging" photos. This is presumably done to allow advertisers to more effectively target individual users - for example, a person identified in a photo with a barbeque may receive adverts for propane gas.

Facial recognition is also enjoying a heavy amount of attention from the security services due to it's use in identifying persons of interest. Case studies have been performed using images released to the public to ascertain how effective facial recognition is when looking for a specific person. In particular, Joshua Klontz and Anil Jain performed a case study using the images of the "Boston Bombers" against a set of test data (Klontz and Jain, 2013). Their approach was successful in recognising one of the perpetrators from a picture taken from his social media account (See appendix B).

### 1.1.5 Constrained vs Unconstrained

While facial recognition software has come a long way, achieving accuracy rates of up to 99% on small, consistent data sets (Best-Rowden et al., 2014), it is still in it's infancy when it comes to identifying people in "unconstrained" images. Images taken in the wild may have large variations in pose, facial occlusion and ambient lighting. This makes it difficult to identify facial features or markers (such as iris distinace, nasal distance, blemishes) which in turn has a negative impact on accuracy rates (Klare et al., 2015). When looking at applications of face recognition software with unconstrained datasets, matches are typically achieved when the test image has similar pose, occlusion and lighting as the sample image (See appendix B).

## 1.2 Theory

### 1.2.1 Prior Work

## 1.3 SOCMINT

### 1.3.1 Prior Knowledge Attacks

### 1.3.2 HUMINT

### 1.3.3 Social Engineering

### 1.3.4 Spearphish

### 1.3.5 Individual vs Group Data

### 1.3.6 Quantity of Information

### 1.3.7 Accessibility of Data

### 1.3.8 Uses

### 1.3.9 Challenges and Constraints

## 1.4 Facial Recognition

### 1.4.1 Current Applications

### 1.4.2 State of the Art

### 1.4.3 Challenges

### 1.4.4 Recent Advances

### 1.4.5 Unconstrained Facial Recognition

# References

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# Appendices

## A  Threat Graph
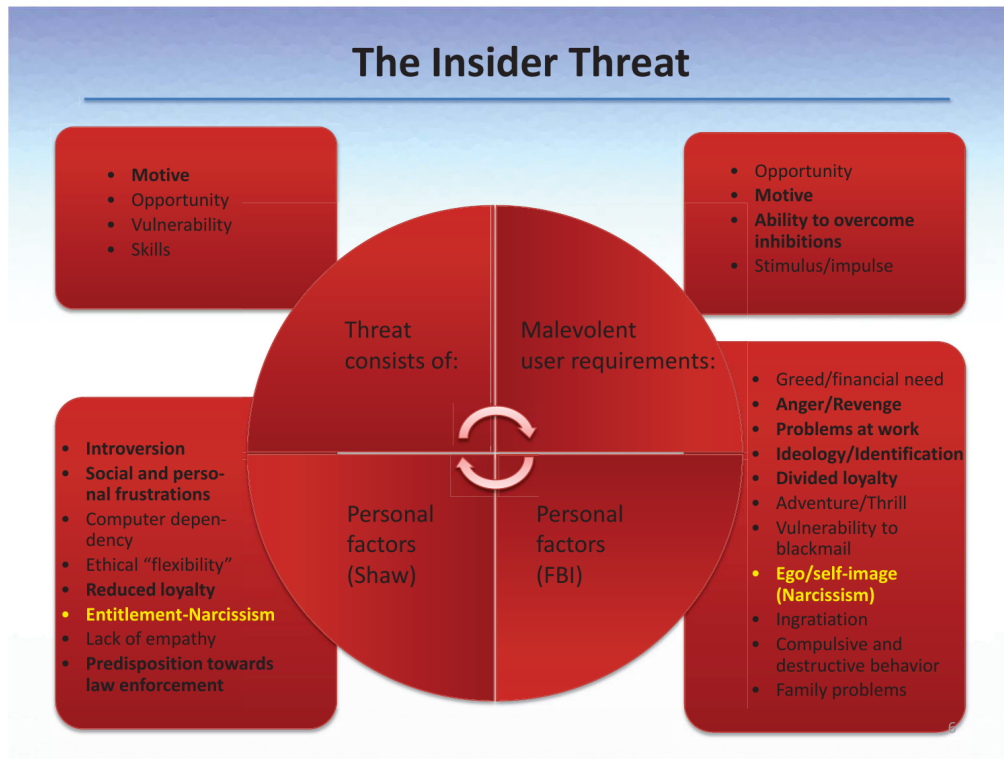


Graph of insider threat factors (Kandias and Stavrou, 2015).
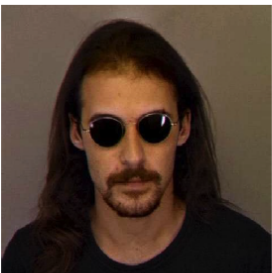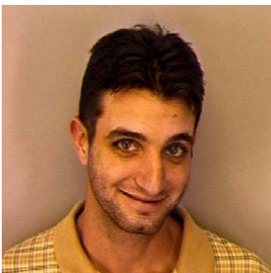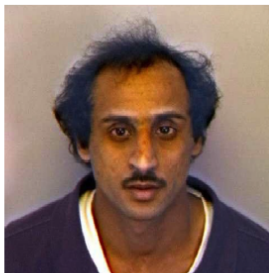
# B  "Boston Bomber" Identification
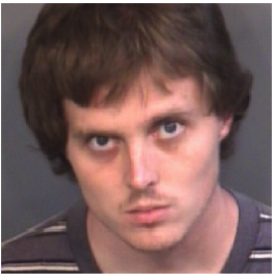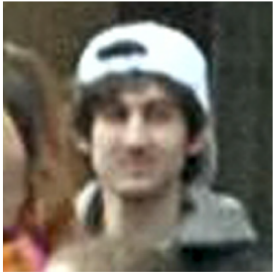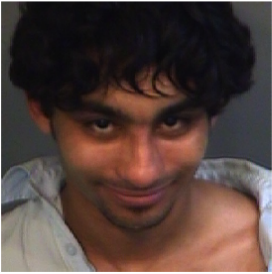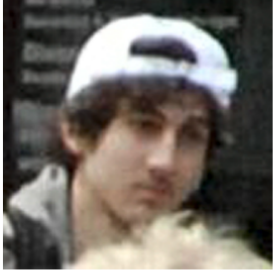
| Probe | Rank 1 | Rank 2 | Rank 3 |
|-------|--------|--------|--------|

Table of potential matches, note the correct identification from the picture taken from social media with similar pose and lighting (Klontz and Jain, 2013).