

Linux Authentication and PAM

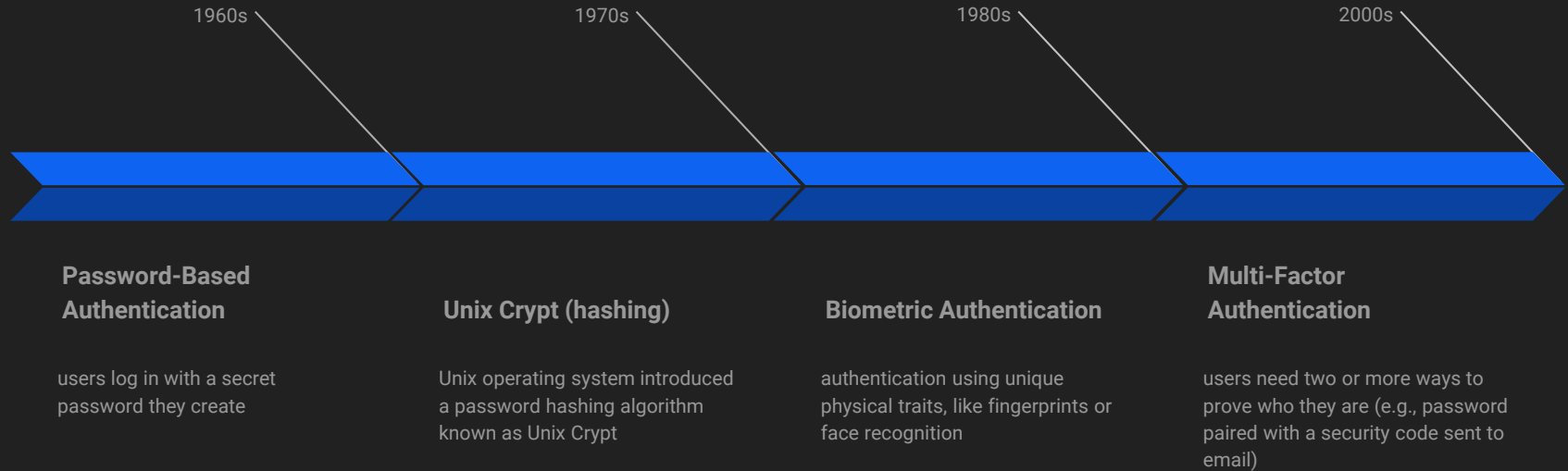
Henry Roeth

What Is Authentication?

- the process of verifying an identity through something someone knows, something they have, or something they are
- goal is to prevent unauthorized access and protect against security threats, such as identity theft, data breaches, and system intrusions

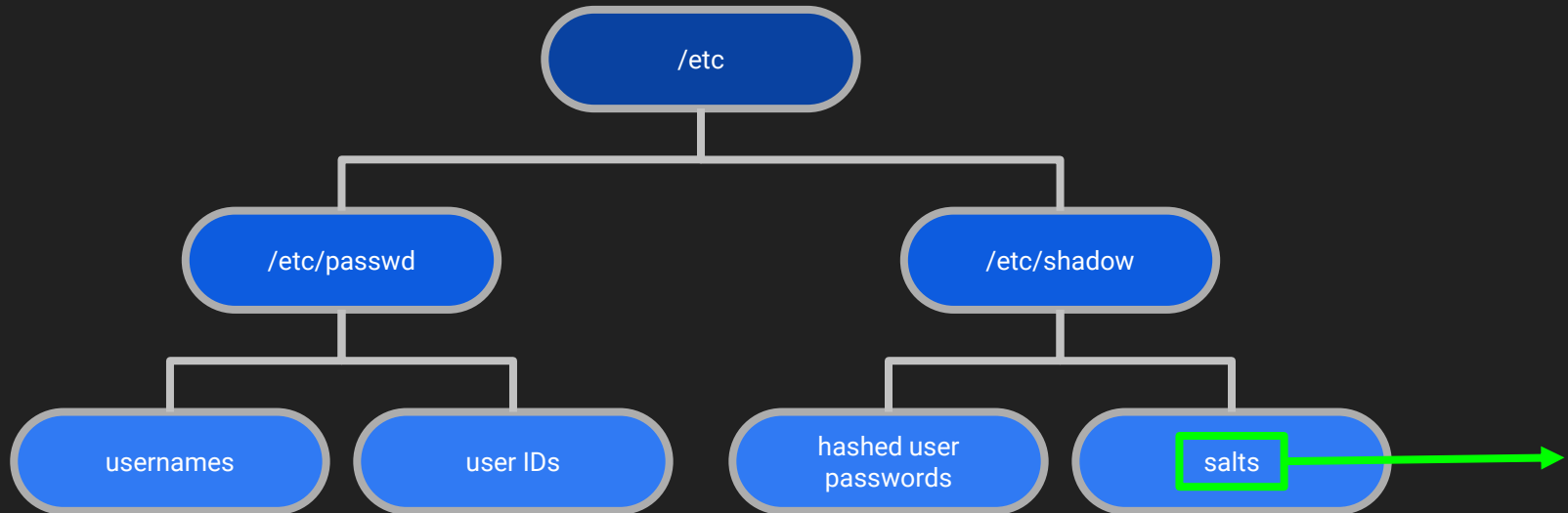


History of Authentication



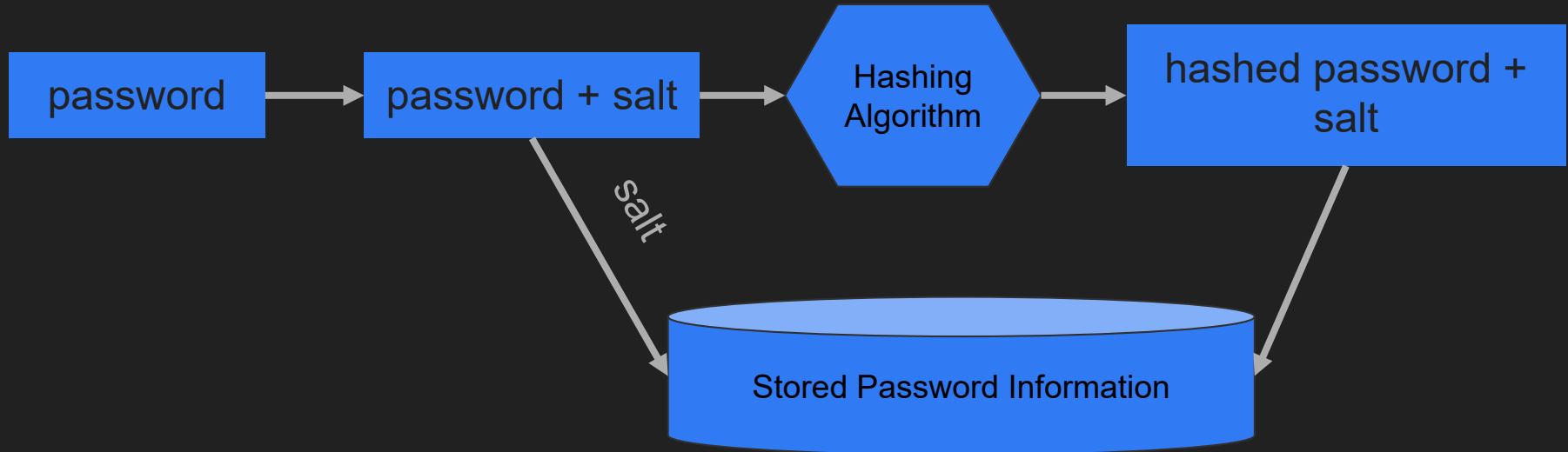
Authentication in Linux

- security information stored in the `/etc/shadow` and `/etc/passwd` files
- `/etc/shadow` file added for extra security as only the root user has inherent access



Salts

- random data that is added to a password before it is hashed
- makes the hashed output unique even if the input is the same as another
- prevents attackers from using precomputed hash tables (rainbow tables)



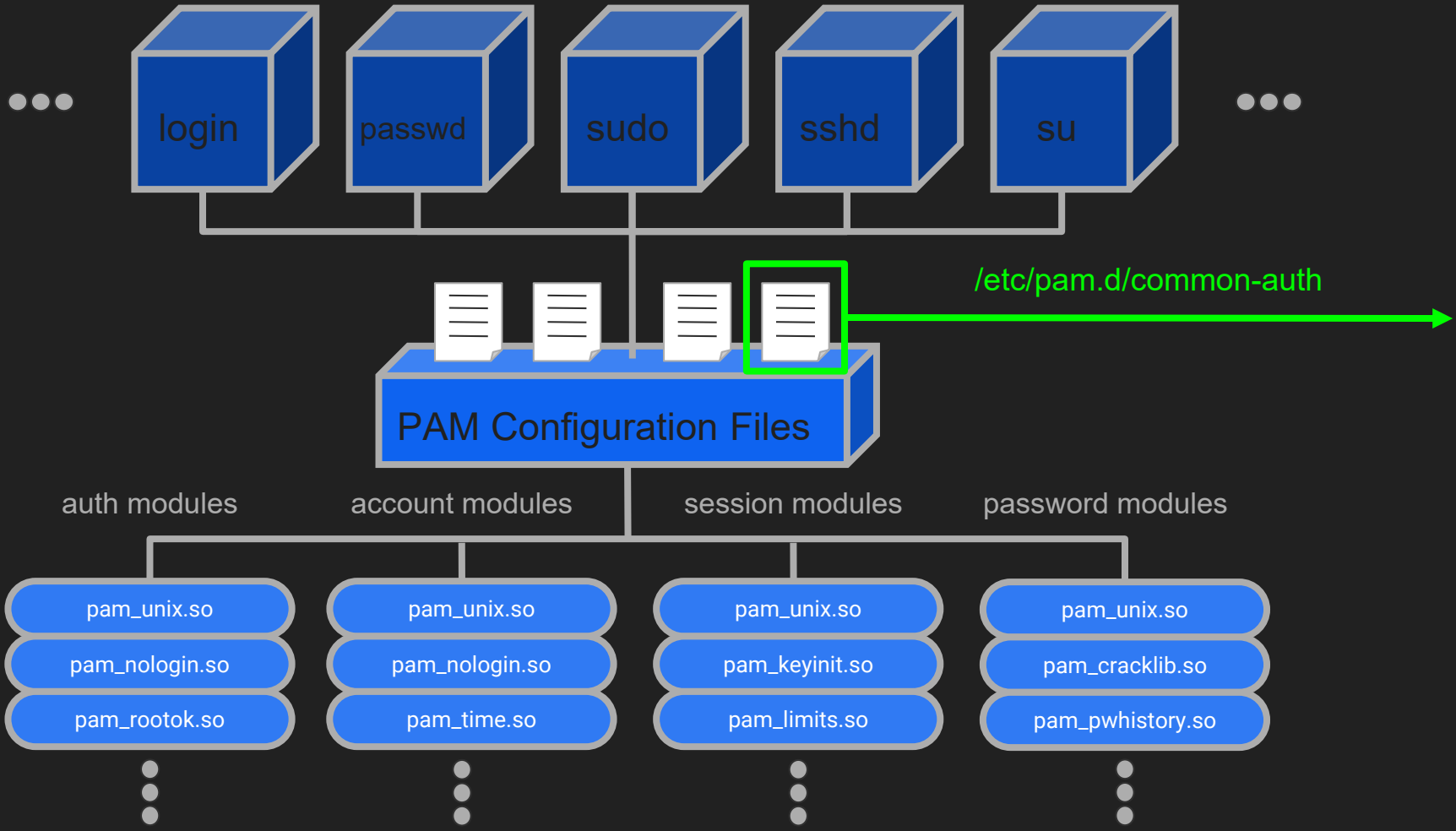
What is PAM?

- PAM (Pluggable Authentication Module) mechanism integrates various low-level authentication schemes into a high-level API, allowing programs that require some form of authentication to be developed independently from the desired authentication scheme

like a multitool



Applications



```
#
# /etc/pam.d/common-auth - authentication settings common to all services
#
# This file is included from other service-specific PAM config files,
# and should contain a list of the authentication modules that define
# the central authentication scheme for use on the system
# (e.g., /etc/shadow, LDAP, Kerberos, etc.). The default is to use the
# traditional Unix authentication mechanisms.
#
# As of pam 1.0.1-6, this file is managed by pam-auth-update by default.
# To take advantage of this, it is recommended that you configure any
# local modules either before or after the default block, and use
# pam-auth-update to manage selection of other modules. See
# pam-auth-update(8) for details.

# here are the per-package modules (the "Primary" block)
auth      [success=2 default=ignore] pam_unix.so nullok
auth      [success=1 default=ignore] pam_sss.so use_first_pass
# here's the fallback if no module succeeds
auth      requisite                    pam_deny.so
# prime the stack with a positive return value if there isn't one already;
# this avoids us returning an error just because nothing sets a success code
# since the modules above will each just jump around
auth      required                    pam_permit.so
# and here are more per-package modules (the "Additional" block)
auth      optional                    pam_cap.so
# end of pam-auth-update config
```


Custom Bypass Module

```
# /etc/pam.d/su  
  
# Allow root to su without passwords  
auth    sufficient pam_rootok.so  
  
# Includes a custom authentication module granting automatic access  
auth    required  pam_always_allow.so
```

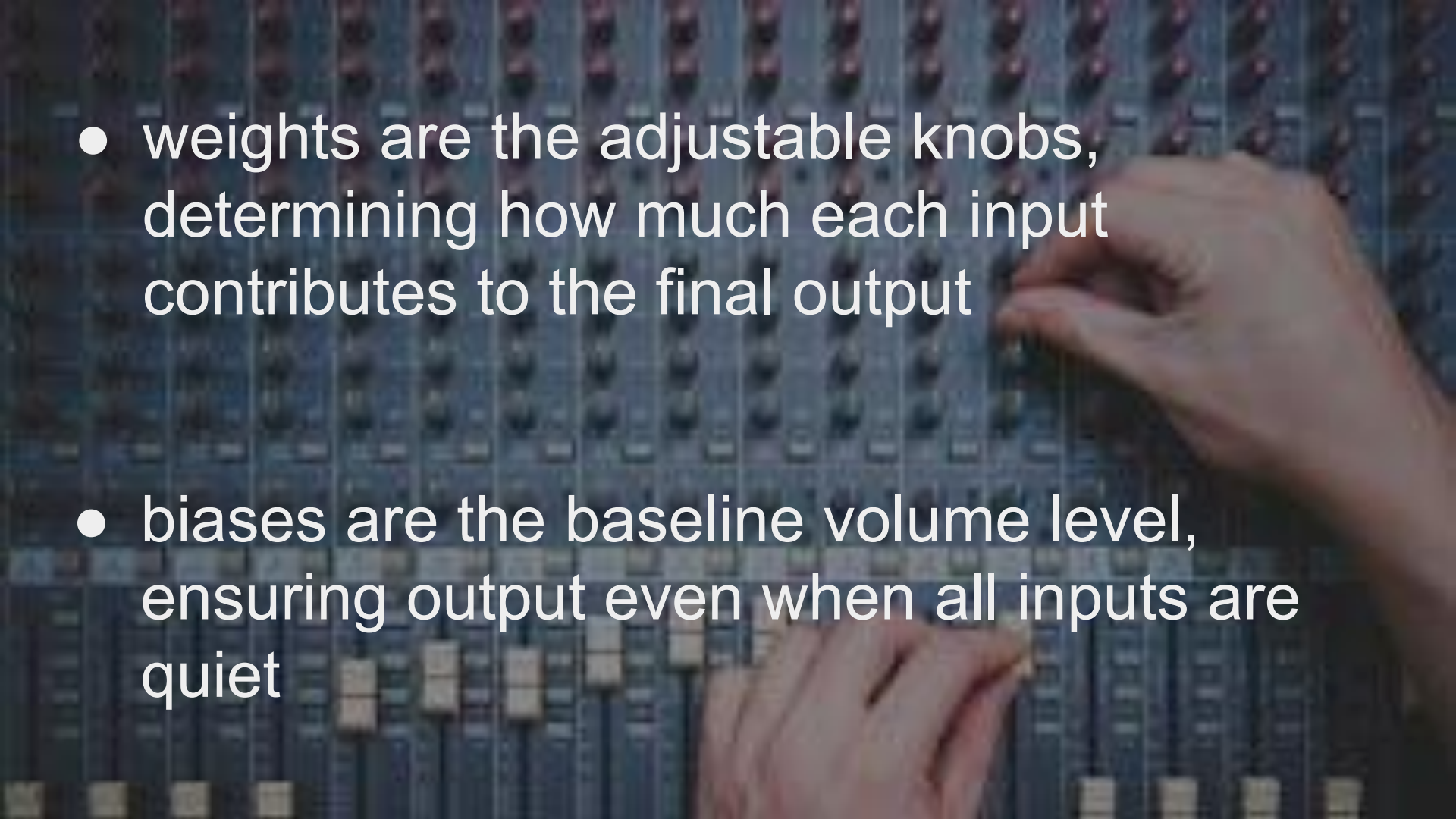
implementation

```
henry@henry-laptop:~$ su test
Automatic access granted!
test@henry-laptop:/home/henry$
```

What Is Machine Learning?

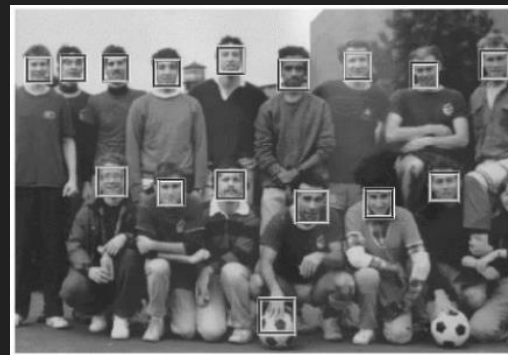
- a subset of artificial intelligence that enables systems to learn from data and make predictions or decisions without being explicitly programmed
- models are trained with “weights” and “biases”
- weights are like parameters that adjust the contribution of different inputs
- biases act as additional values that allow the model to account for situations where all inputs are zero or have no effect

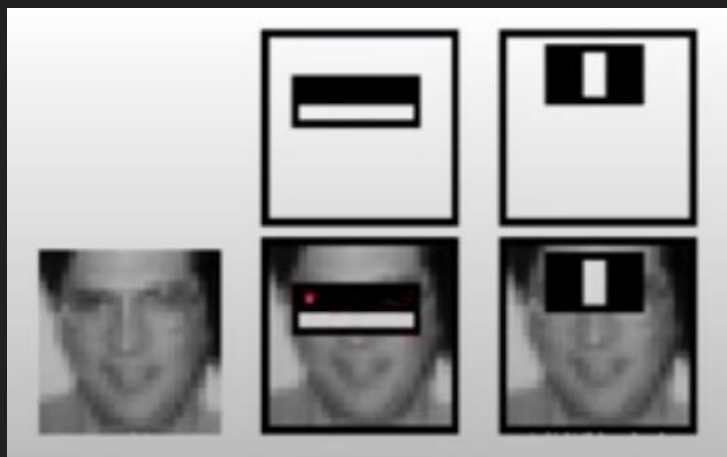
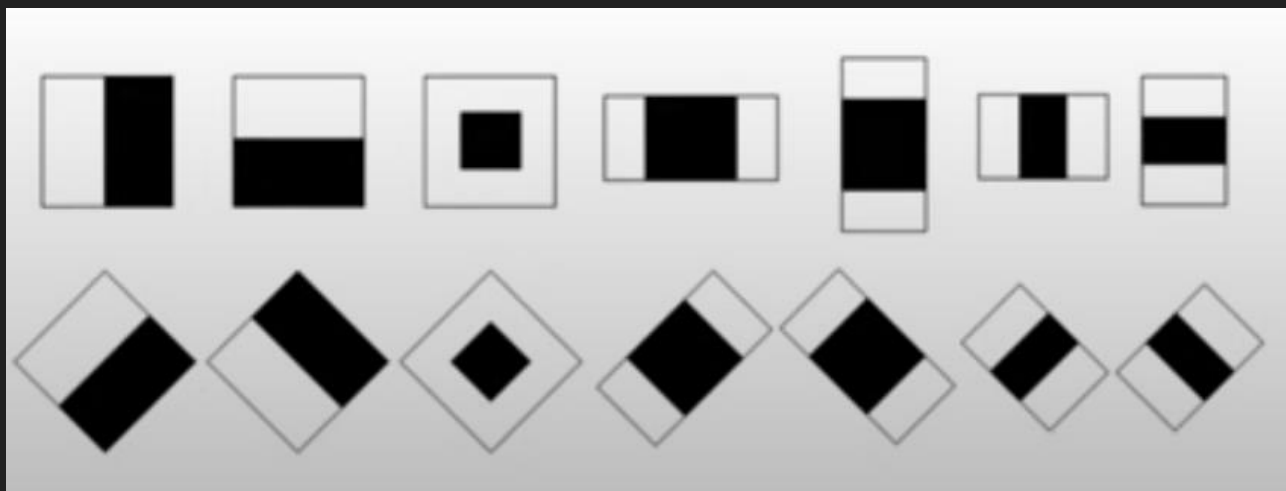
think of a sound mixer...

- 
- A close-up, slightly blurred photograph of a hand adjusting a knob on a vintage audio mixing console. The console has many rows of similar knobs and sliders, creating a repetitive pattern. The lighting is soft, and the colors are muted, giving it a professional, technical feel.
- weights are the adjustable knobs, determining how much each input contributes to the final output
 - biases are the baseline volume level, ensuring output even when all inputs are quiet

Haar Cascade Classifier

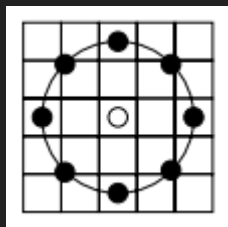
- named after Alfred Haar, created by Viola and Jones in 2001
- detects simple rectangular patterns that capture contrast differences in the image, such as edges or texture variations
- trained with a large dataset of positive (containing object of interest) and negative (not containing object of interest) images
- algorithm learns to distinguish between positives and negatives



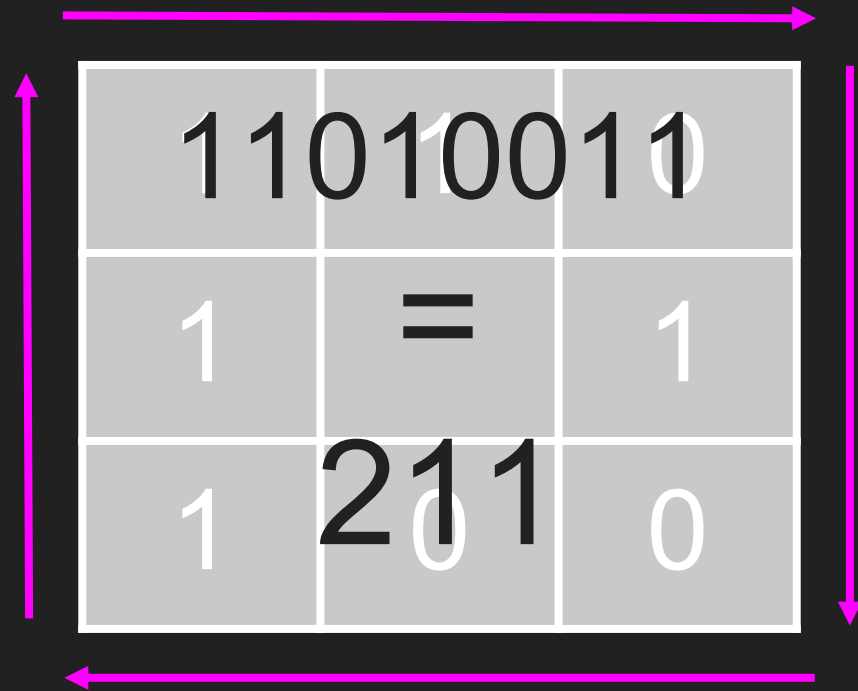
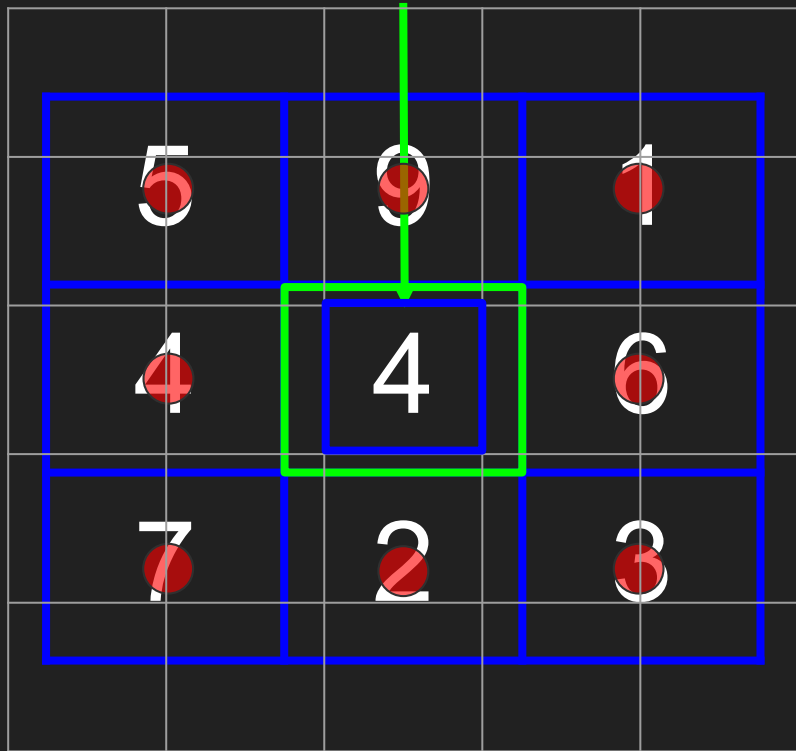


Local Binary Patterns Histograms (LBPH)

- feature-based face recognition method using texture classification
- LBPs encode local texture patterns by comparing each pixel with its neighbors
- binary values are calculated about a center pixel (threshold)
 - values \geq threshold are assigned a binary value of 1
 - values $<$ threshold are assigned a binary value of 0
- these “blocks” (9x9 pixel area) yield a decimal value using the binary values
- edges and features can be extracted using these values and analyzed in histograms showing the frequency of specific block values (0-255)

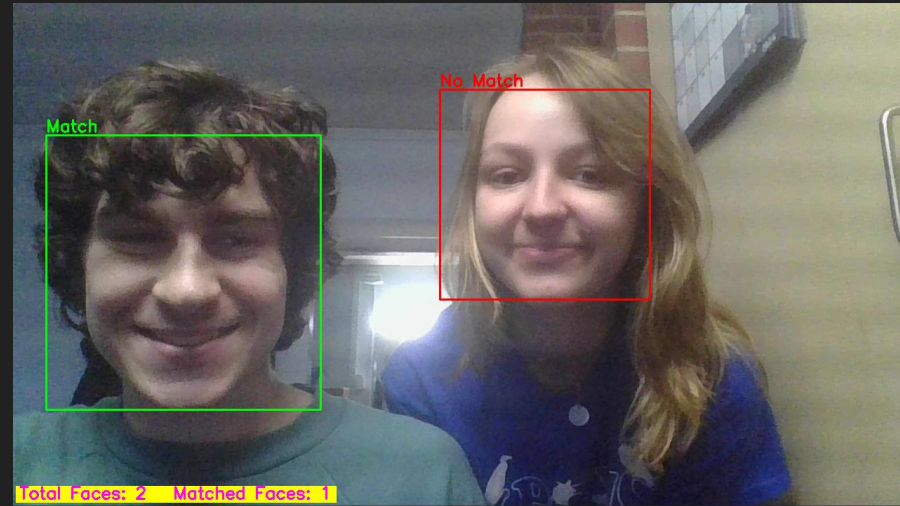


threshold



Results

- face recognition performed pretty well with minimal training data
- able to differentiate matches from non-matches
- working face recognition logic, but runtime security error causing PAM to default to a failure
- logic performed with a pre-captured image yielded a success
- camera usage causing failure



```
henry@henry-laptop:~$ su test
Please hold still. Recording will begin in 1 seconds.
Recording started. Please hold still.
Average confidence: 16.7201
SUCCESS!
Password:
su: Authentication failure
```

```
henry@henry-laptop:~$ su test
SUCCESS!
test@henry-laptop:/home/henry$
```

LIVE DEMO!

Improvements

- train LBPH model with more gallery images of varying distances from camera, shadows, head positions, etc.
- train with data on multiple individuals to identify more than one person
- ensure there are no runtime errors arising when using the camera

References

- Ahonen, T., Hadid, A., and Pietikäinen, M. (2004). Face recognition with local binary patterns.
- Geisshirt, K. (2007). Pluggable Authentication Modules: The Definitive Guide to PAM for Linux Sysadmins and C Developers. Packt Publishing, Birmingham, UK.
- Viola, P. and Jones, M. J. (2001). Rapid object detection using a boosted cascade of simple features. Proceedings of the 2001 IEEE Computer Society Conference on Computer Vision and Pattern Recognition. CVPR 2001, 1:1–511.

Questions?