





# How we scaled our deep learning infrastructure to process 1000 photos / second while cutting our AWS bill by 10%

Rino Montiel, VP Engineering







## EyeEm is a photography company that builds the world's leading computer vision technology to connect its global creative community with iconic brands.



# What we've automated

Photo quality rating Popular Auto tagging Model release detection Keywording Captioning Search Personalized aesthetics Concept training platform





# Understanding photography





## Man in Abandoned Building

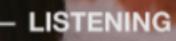


### EyeEm

INDOORS

BRICK WALL

WIRELESS TECHNOLOGY -



CHEERFUL

100

SMARTPHONE

YOUNG WOMAN

HOLDING



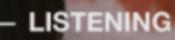
### EyeEm

INDOORS

BRICK WALL

### 89% Aesthetic Score

WIRELESS TECHNOLOGY



CHEERFUL

SMARTPHONE

and the second sec

YOUNG WOMAN

IOLDING





# Three Parts Research, Production, Data





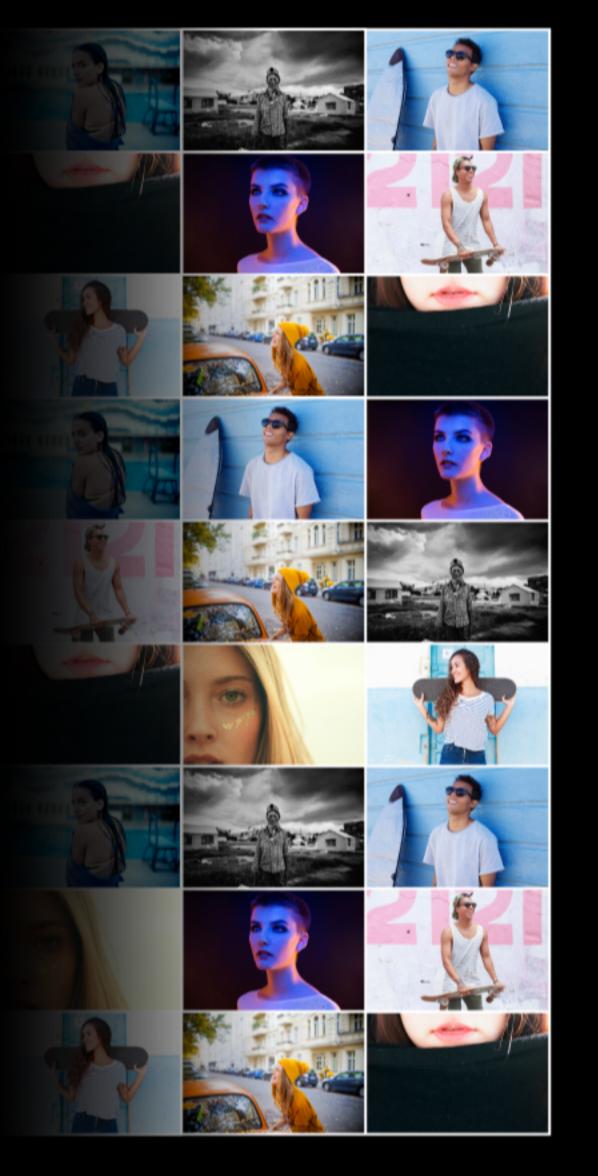
# Beginning: Research

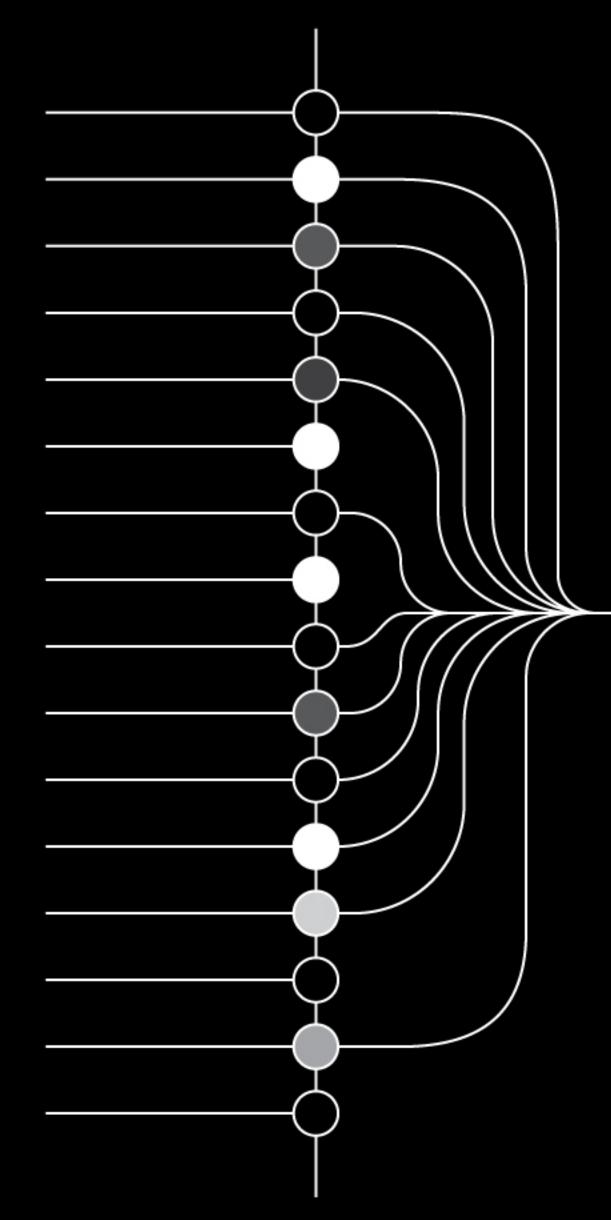
Where the models (and some trouble) come from

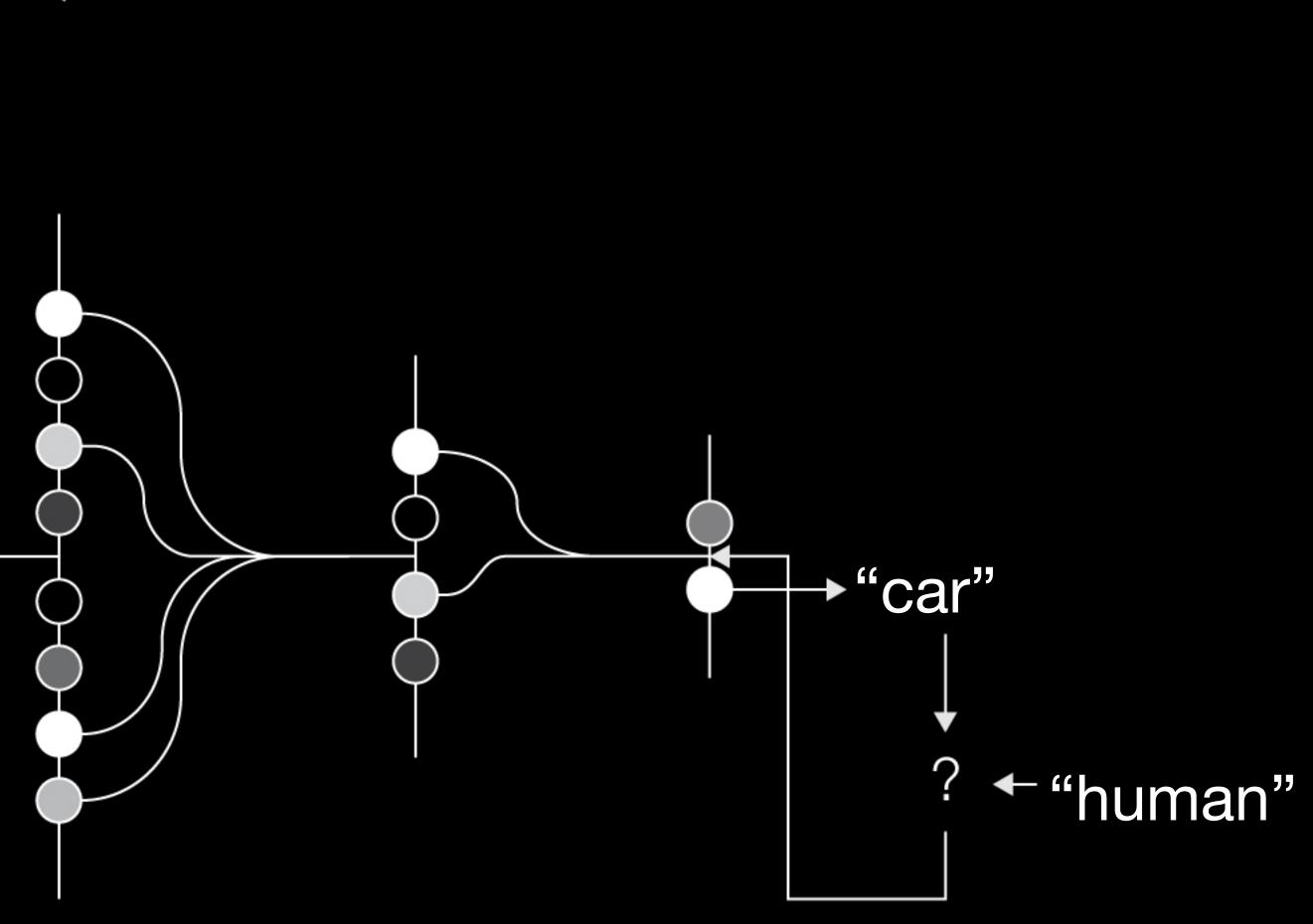


# Training and inference

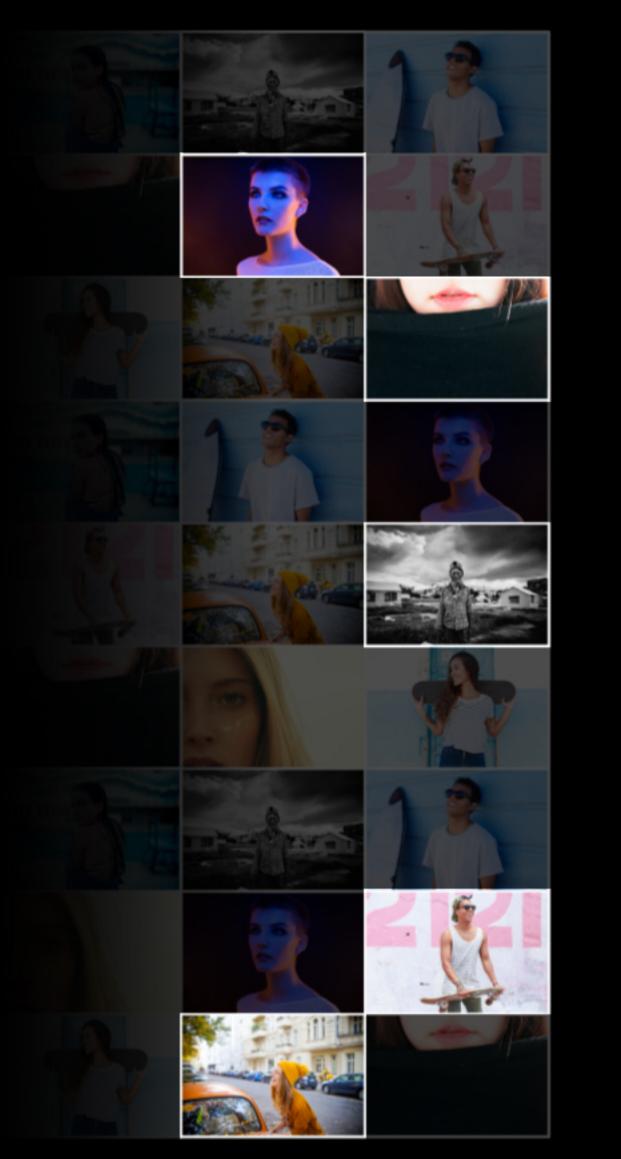


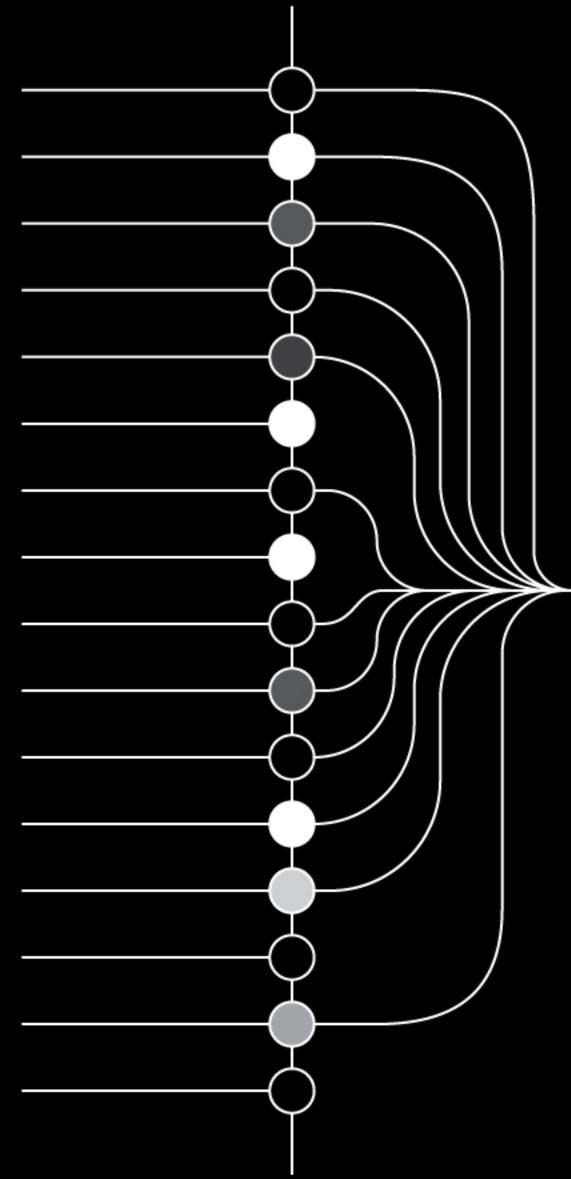


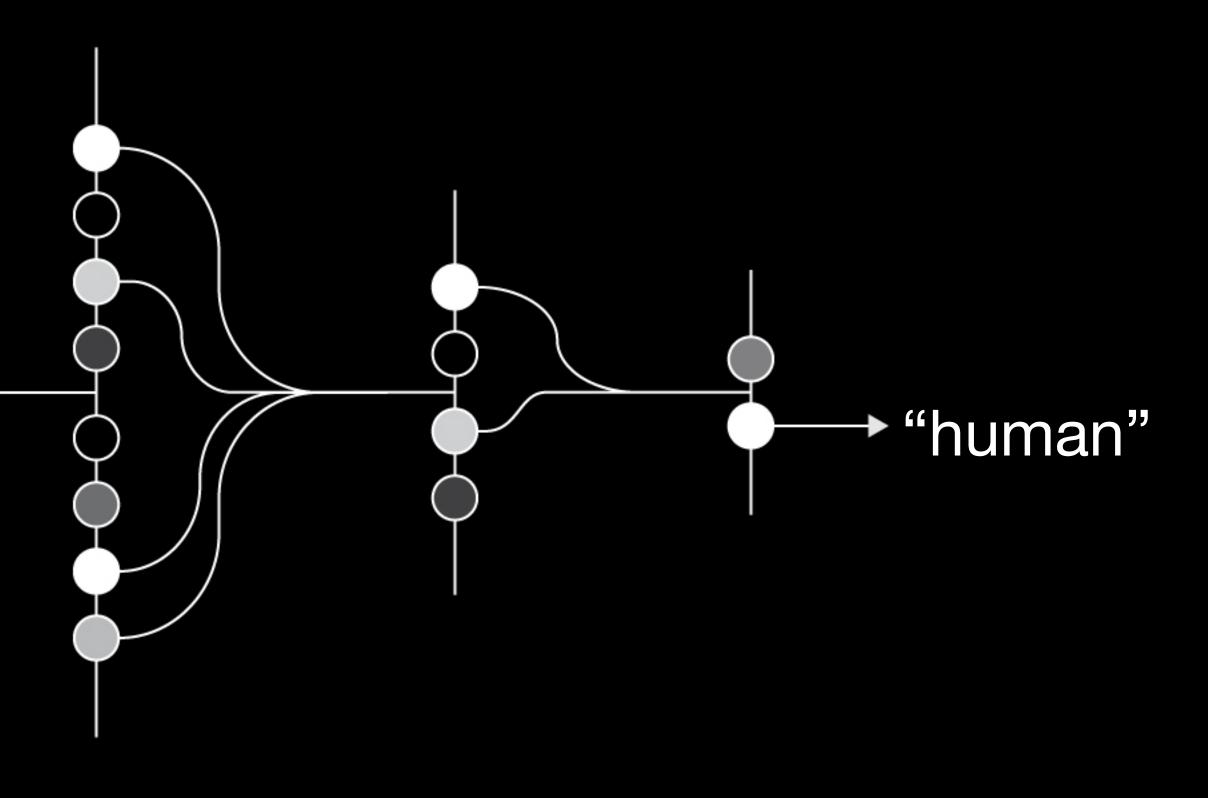














# Deep learning and GPUs

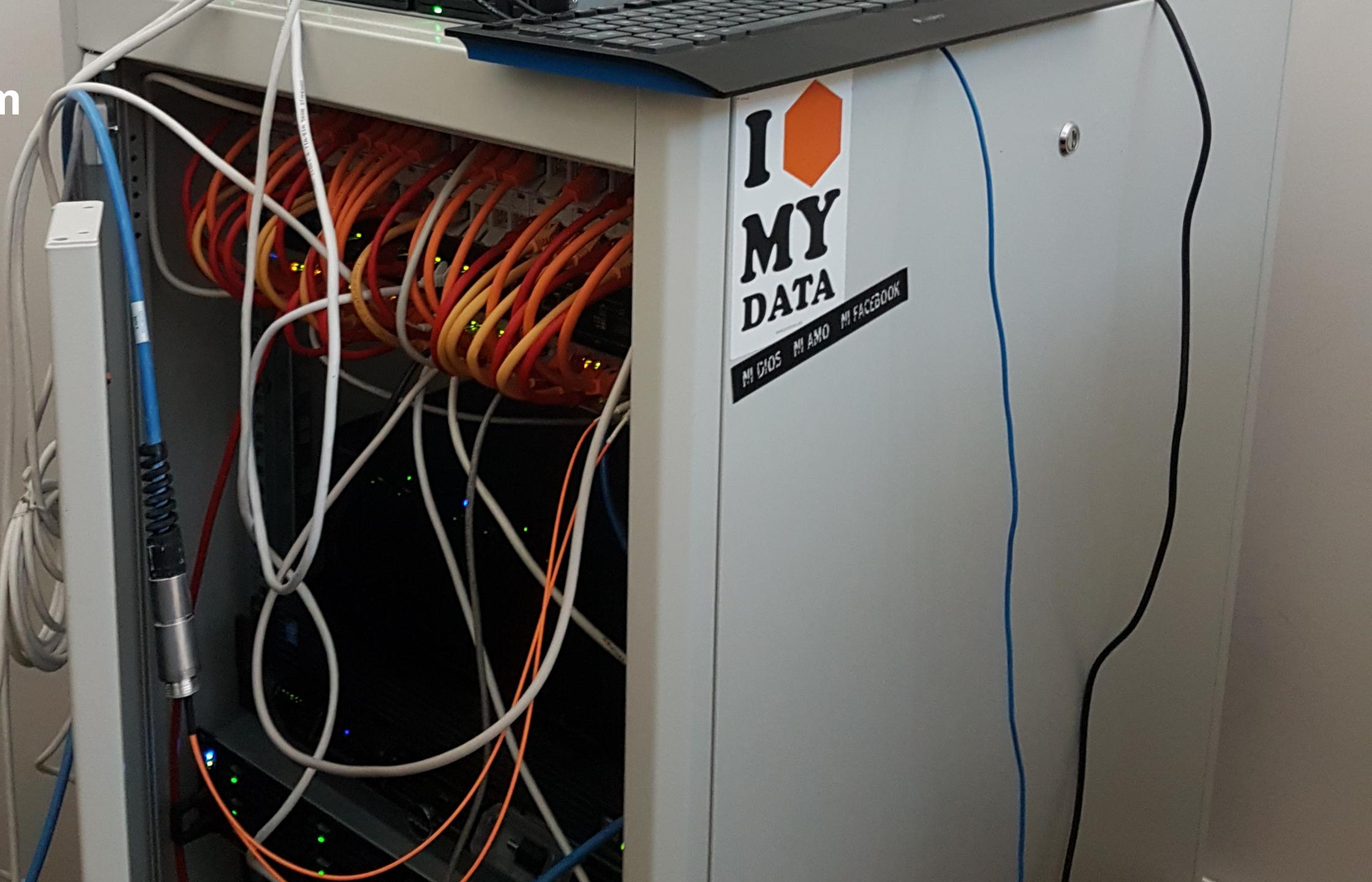
GPUs run deep learning training and inference very efficiently



# Training with millions of photos needs GPUs

More data, more GPUs

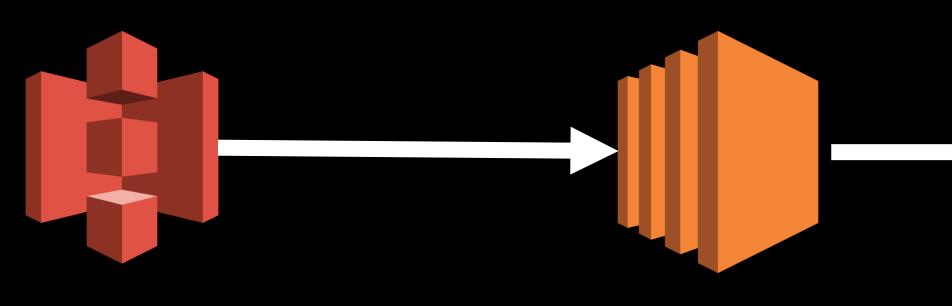
### EyeEm





# Model trained and production ready





### **Thumbnail Service**



### Deep Learning Service





# Problem: Input size changes



# Problem: Resizing algorithm



# Problem! Thumbnail Service

Scaling thumbnail generation for training



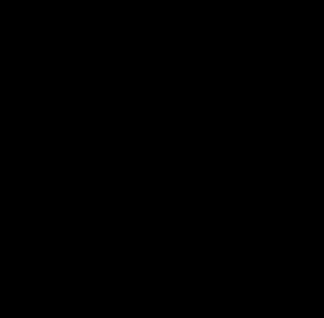
# Problem Thumbnail Service

Scaling thumbnail generation for training

#### **Thumbnail Service**



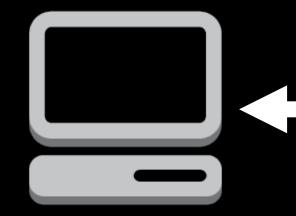












# Problem Thumbnail Service

Scaling thumbnail generation for training

#### Thumbnail Internal Thumbnail Service









Training: Learning new functionality from past data Inference: Applying functionality to new data

Input size can change Resizing algorithm affects performance

Thumbnail generation is load intensive Rely on a mix of on-demand and spot instances \$

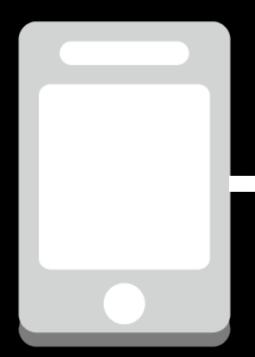


# Middle: Production

Dealing with scale



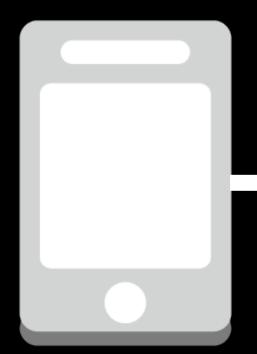














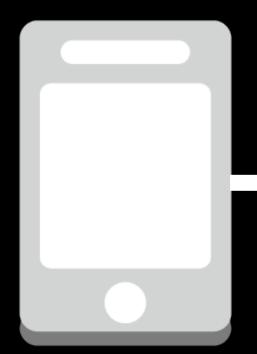


#### Search

#### **Deep Learning Service**

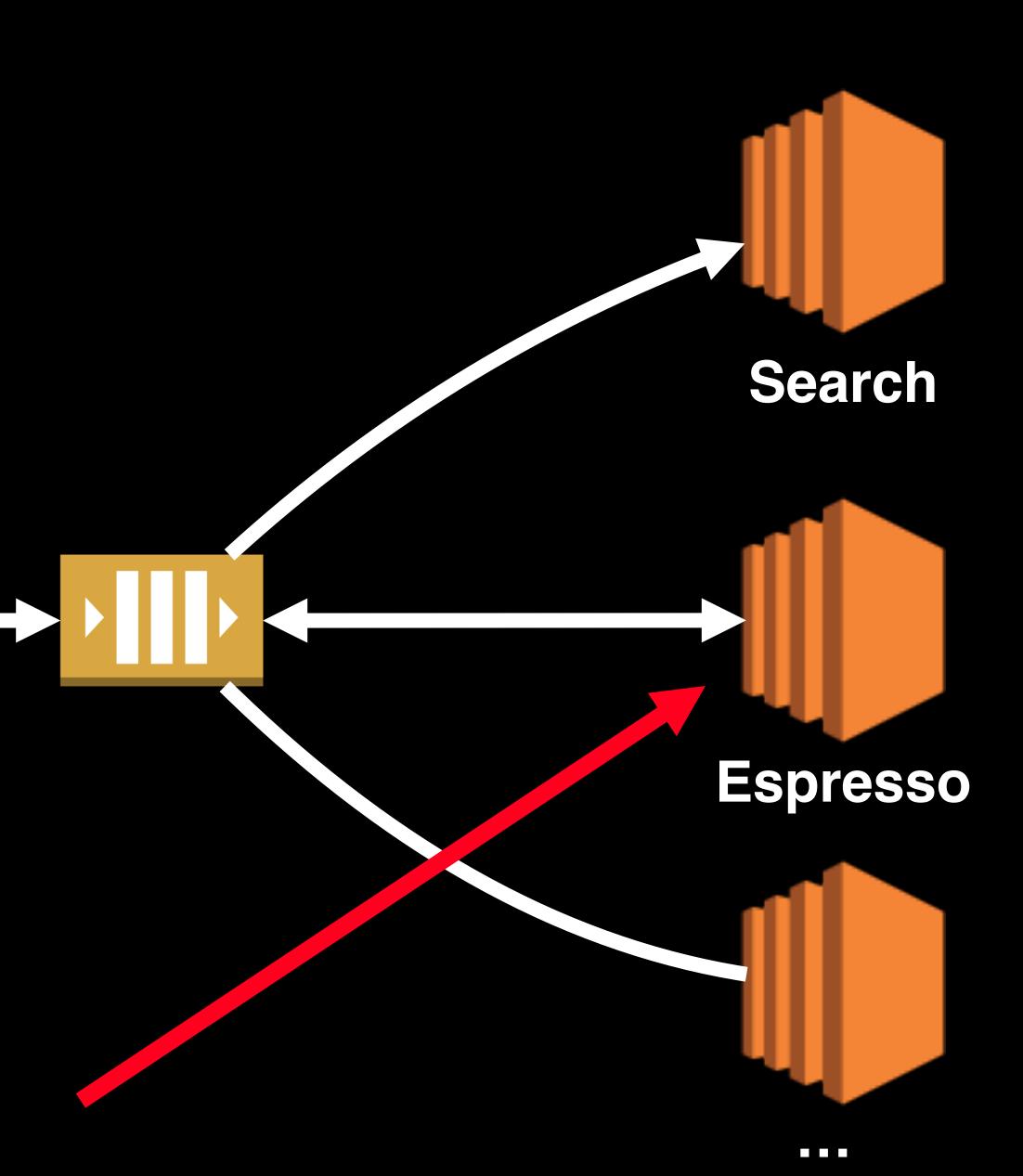




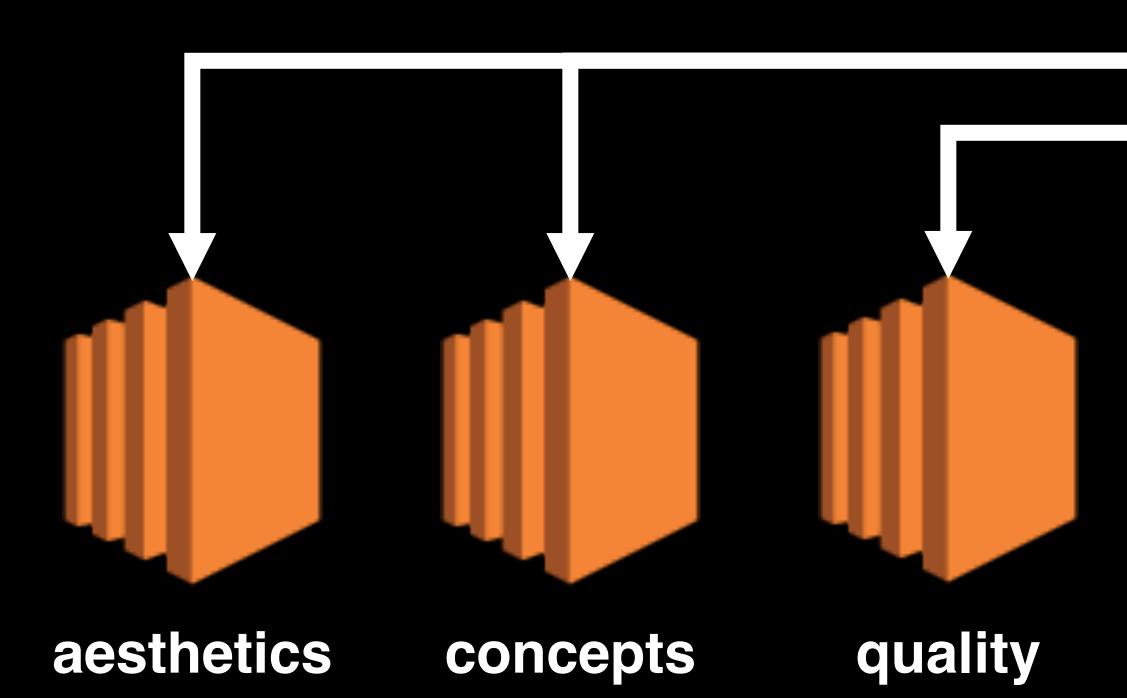












#### Espresso

### G2 machines

#### captions

personalised aesthetics





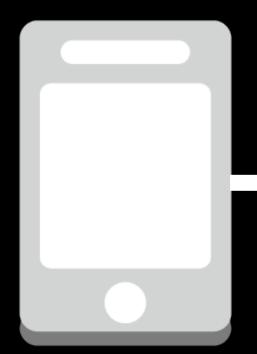


# What we've automated







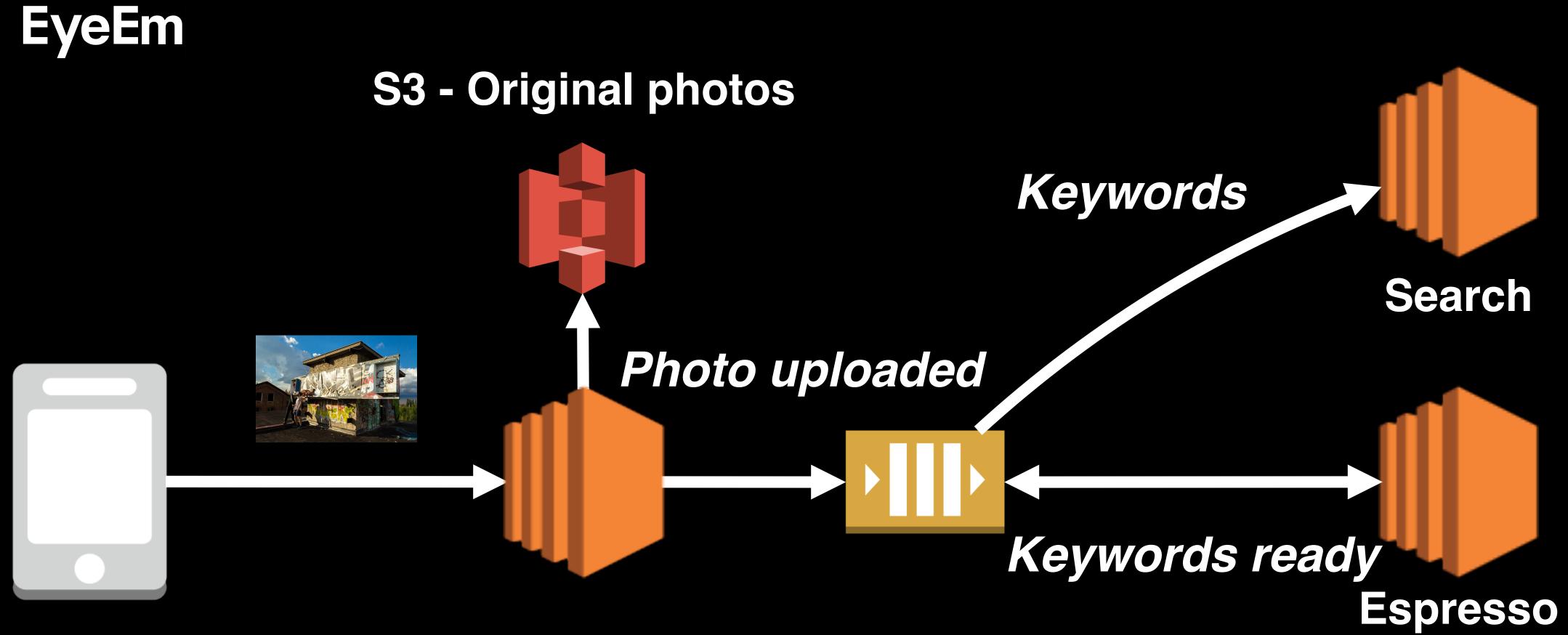












Edge

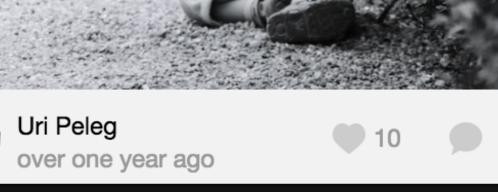
EyeEm	Following	happiness	Q	
Hide filters	- 15	3,612 results		
Market Photos on Eye	– Em Market			
Format   Portrait   Landscape   Square   Widescreen   Panorama		Frik Nardini   about one year	112	
Size Large Medium				

#### Need help?

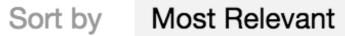
Small

Our photo curators can help you find what you're looking for - free of charge.

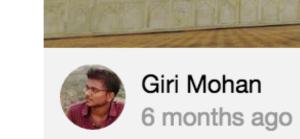
80



Market Dashboard



Madeline Yates 35



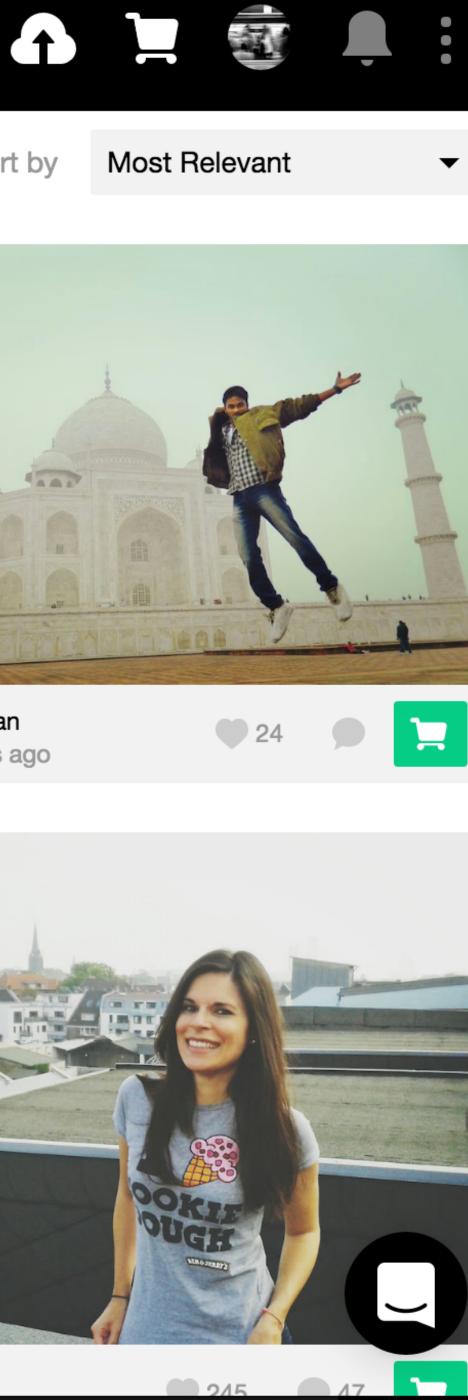




----



11 months ago

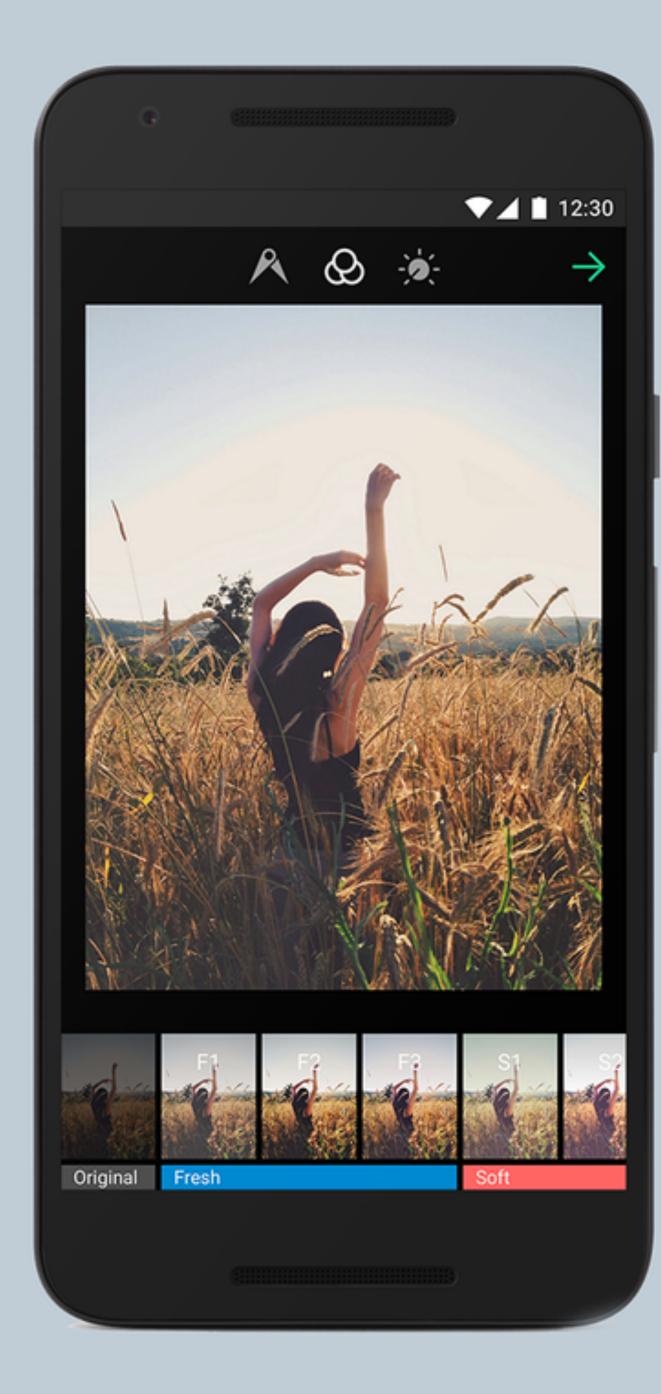


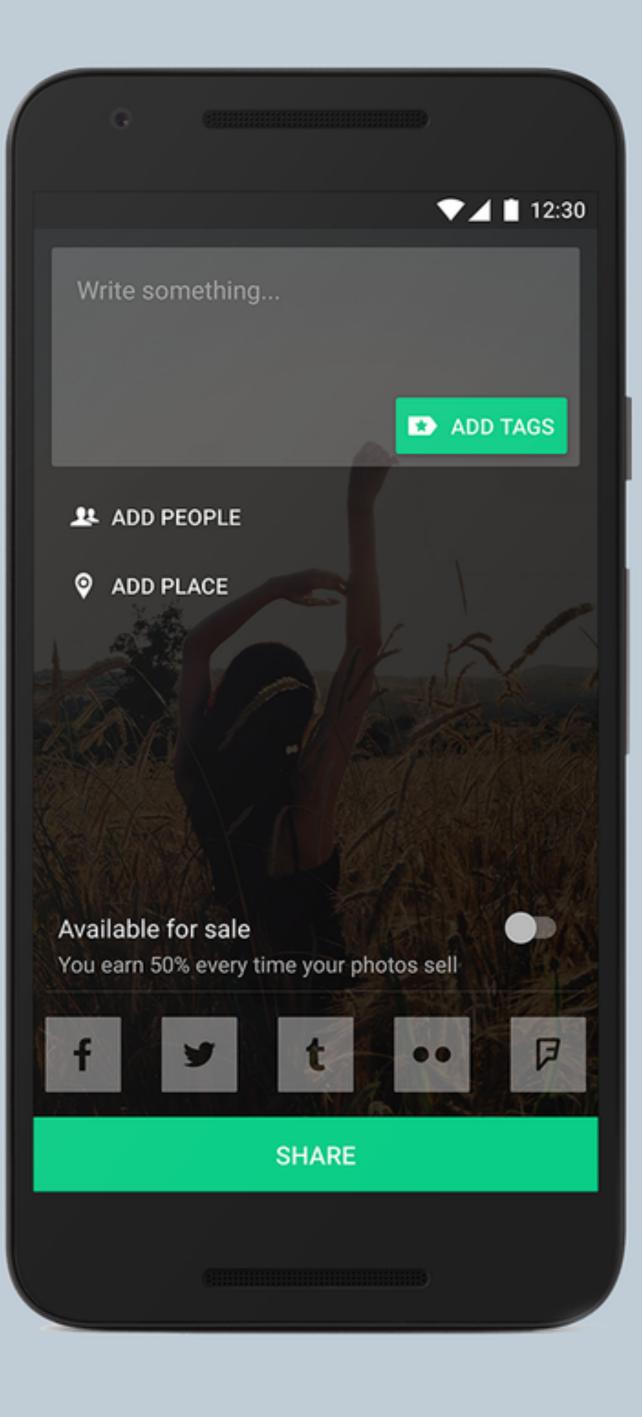


# What we've automated

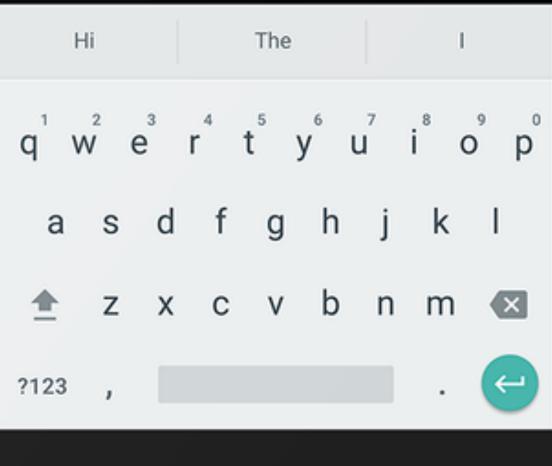
## Auto-tagging







► 12:30 Suggested tags by EyeEm Vision ADD ALL growth field grass plant crop nature close-up farmland focus on foreground tranquility clear sky agriculture rural scene cereal plant farm tranquil scene cultivated land day wheat







# Problem: Latency is too high



# Problem: This is getting expensive...



## Batch inference

"GPU-Based Deep Learning Inference: A Performance and Power Analysis", NVIDIA 2015





#### Batch of base64 encoded images

Array of concepts and aesthetic scores







# Image reviewing pipeline



# The traditional stock photography industry.

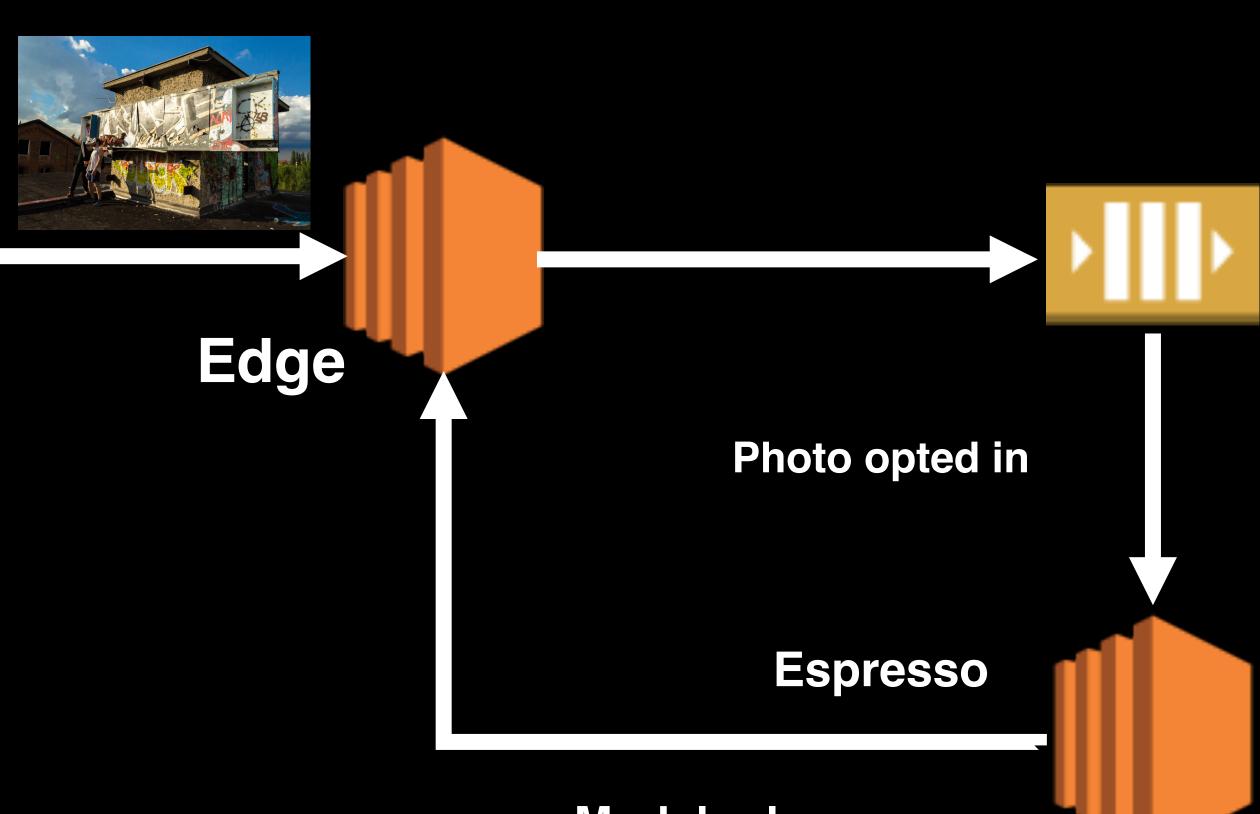




### What we've automated Photo quality rating Model release detection

# Photo reviewing Process





Model releases Quality **Keywords and captions** 



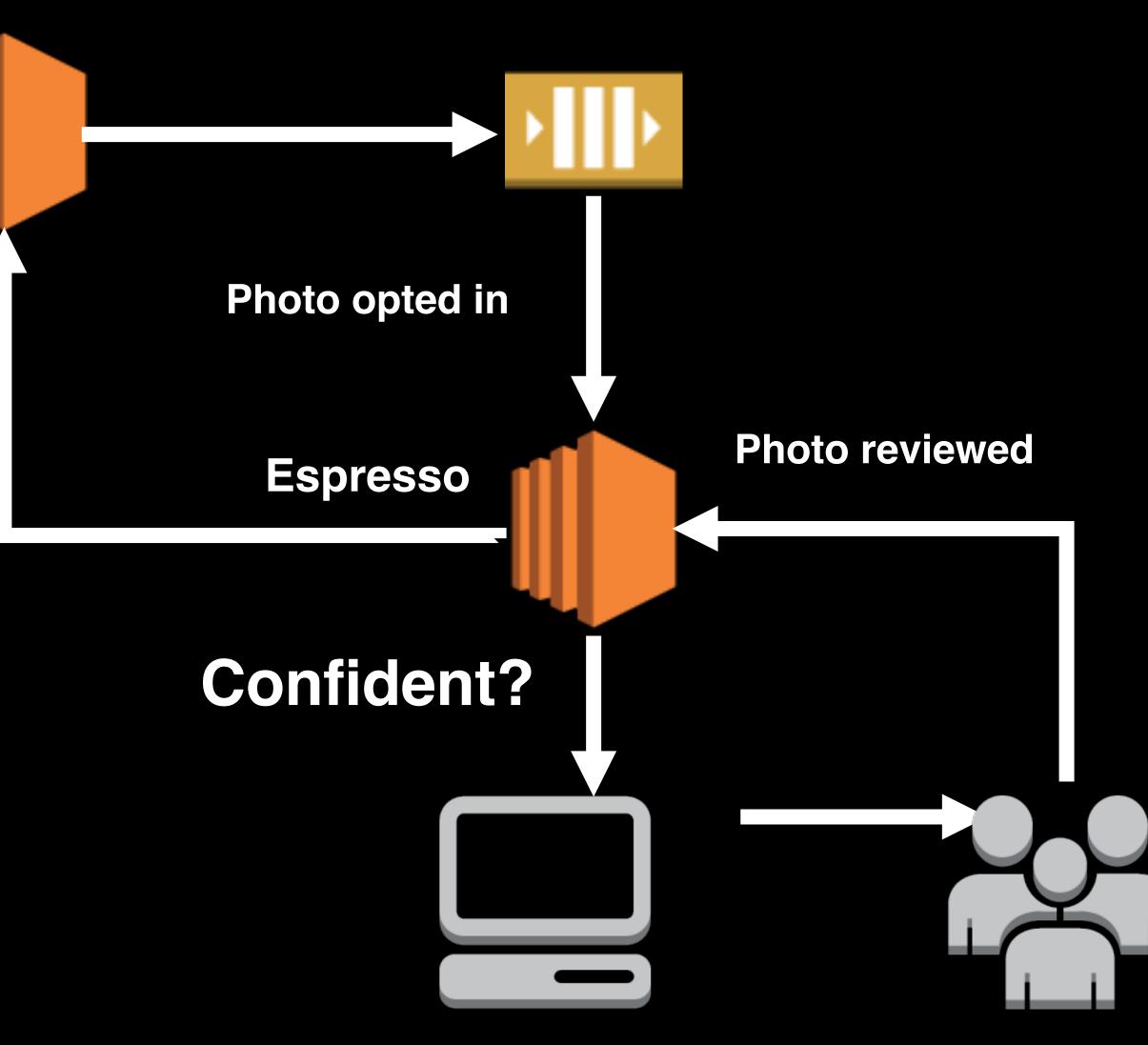
#### Woman photographing through smartphone

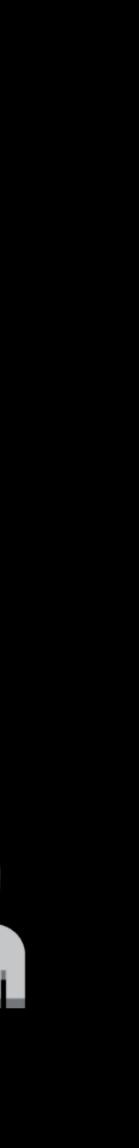




Model releases Quality Keywords and captions

# Photo reviewing process







#### Asynchronous architectures are necessary to scale

#### Batch inference is essential for cost optimization and latency

#### 90% accuracy can be misleading, but it can be optimized





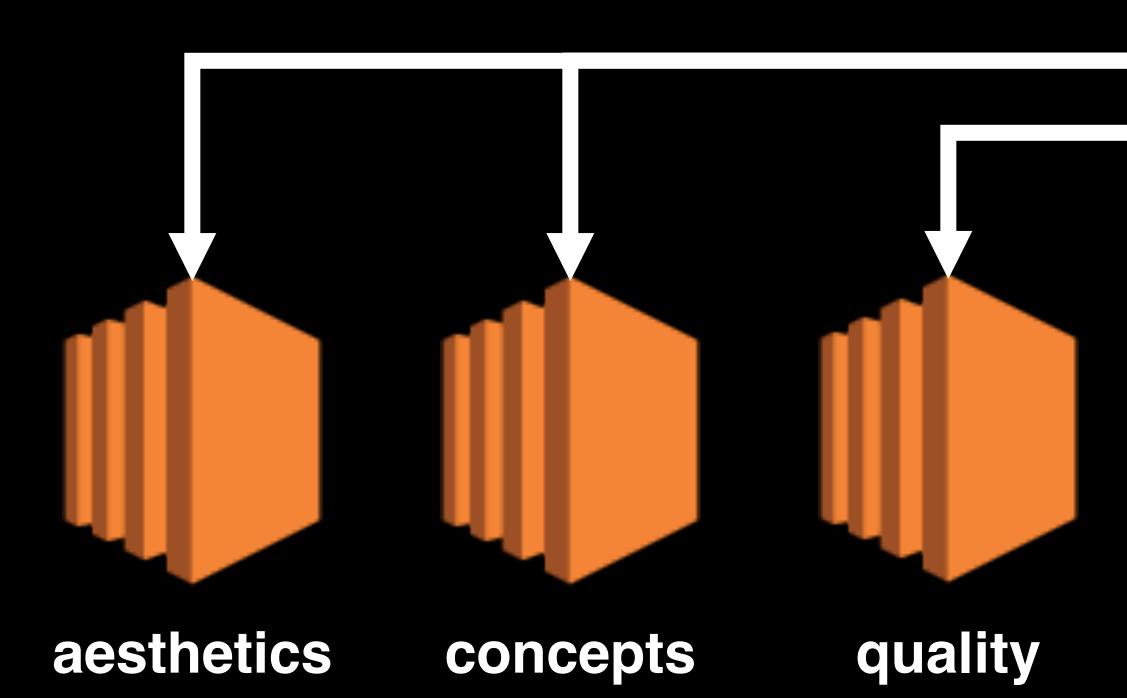
# Everything runs



# Data

Data pipeline for analytics and machine learning





#### Espresso

#### captions

#### personalised aesthetics





# Problem: Too much data!



### One photo

Keywords Captions Confidence Quality Model releases Aesthetics





{"photoId":"102394173","concepts":[["real yles", 0.84903037548065186], ["two people", 0.82873058319091797], ["leisure ",0.78834164142608643],["men",0.78776693344116211],["full length", 0.78717267513275146], ["beauty in 593142145

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#### **Concepts/Confidence**



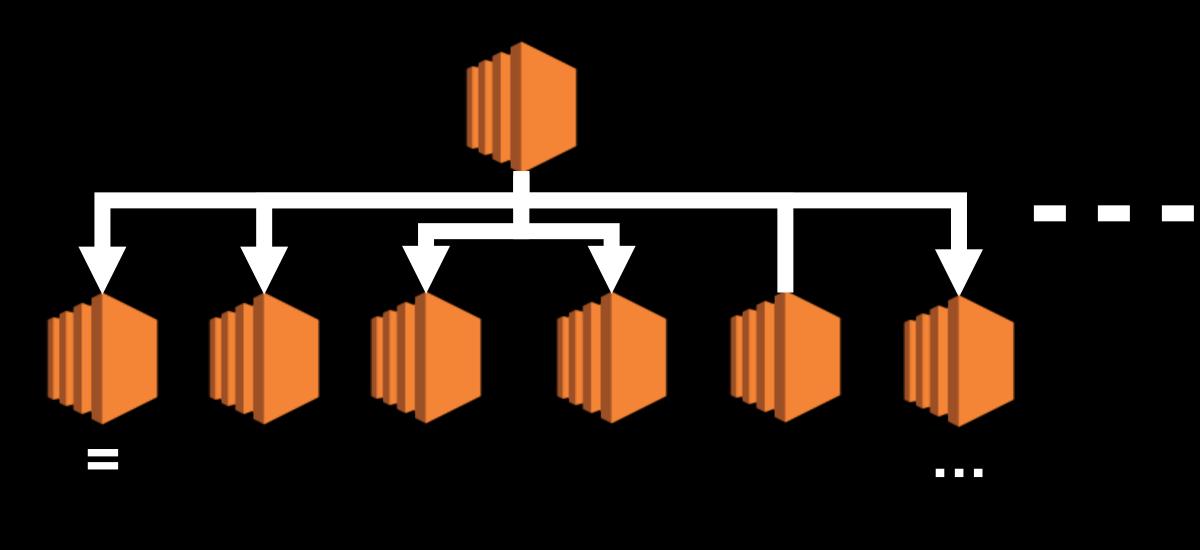
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# Billions of events

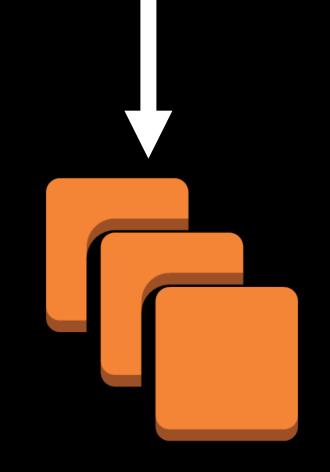






#### Report





#### HDFS/Spark



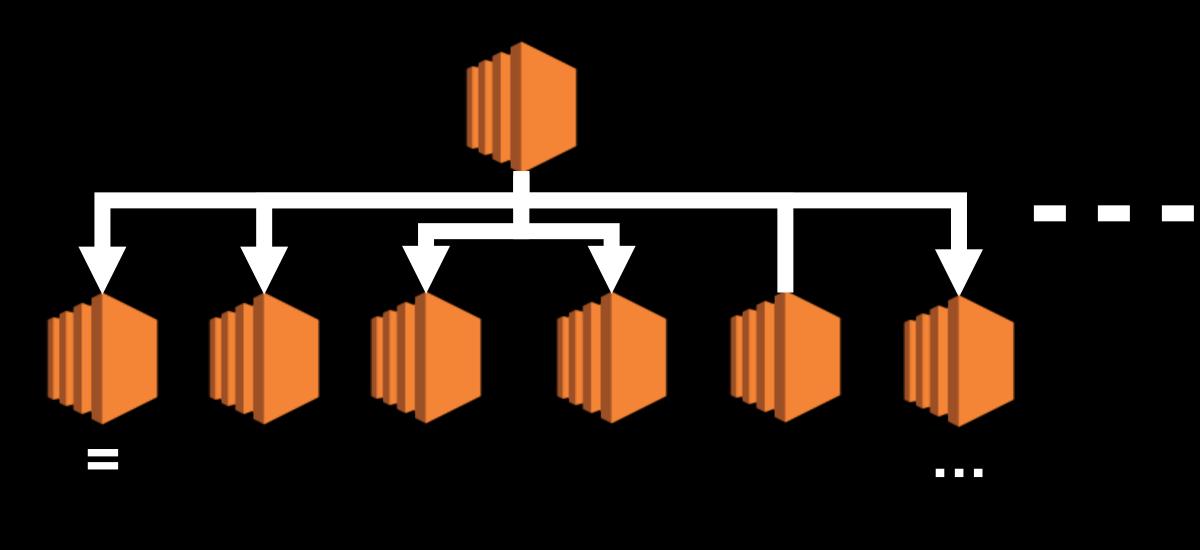


### Our move to EMR



# Analytics pipeline: Spark/EMR/Redshift

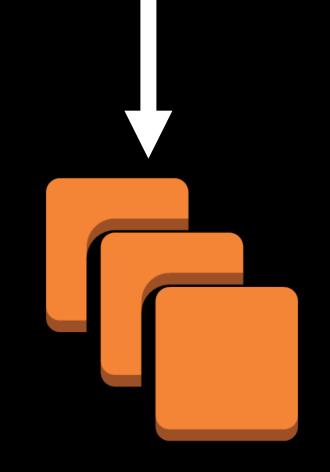






#### Report

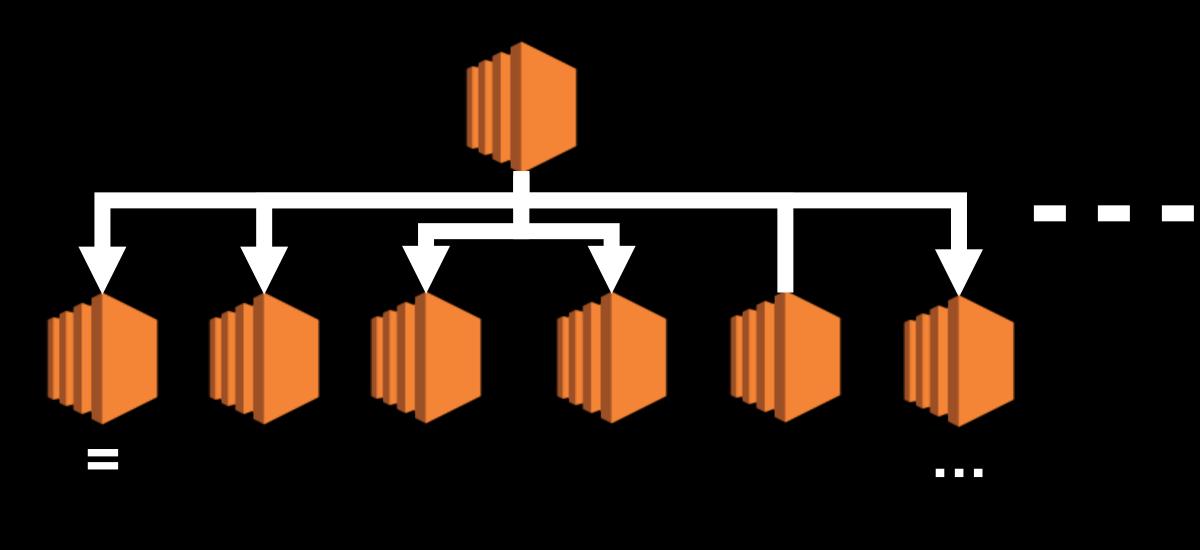




#### HDFS/Spark

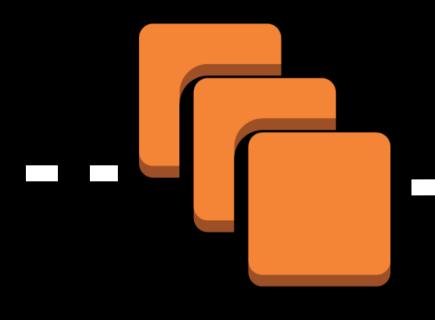




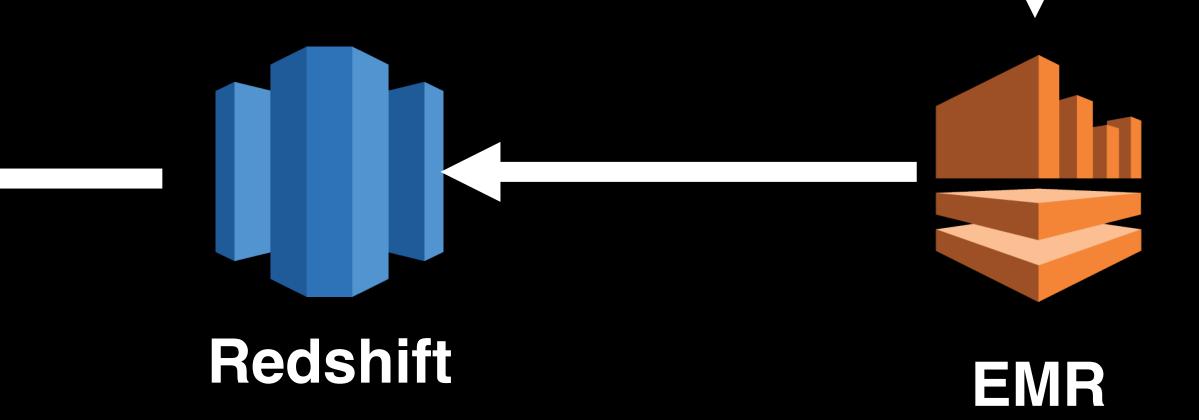




#### Report



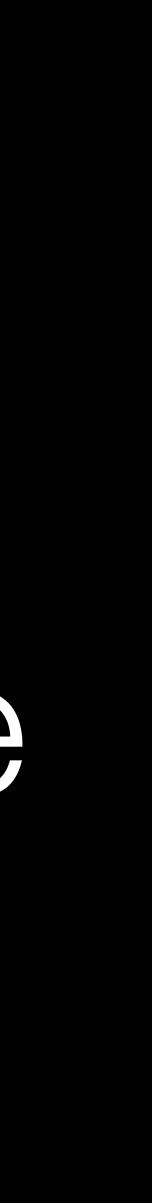
#### **Apache Flume**



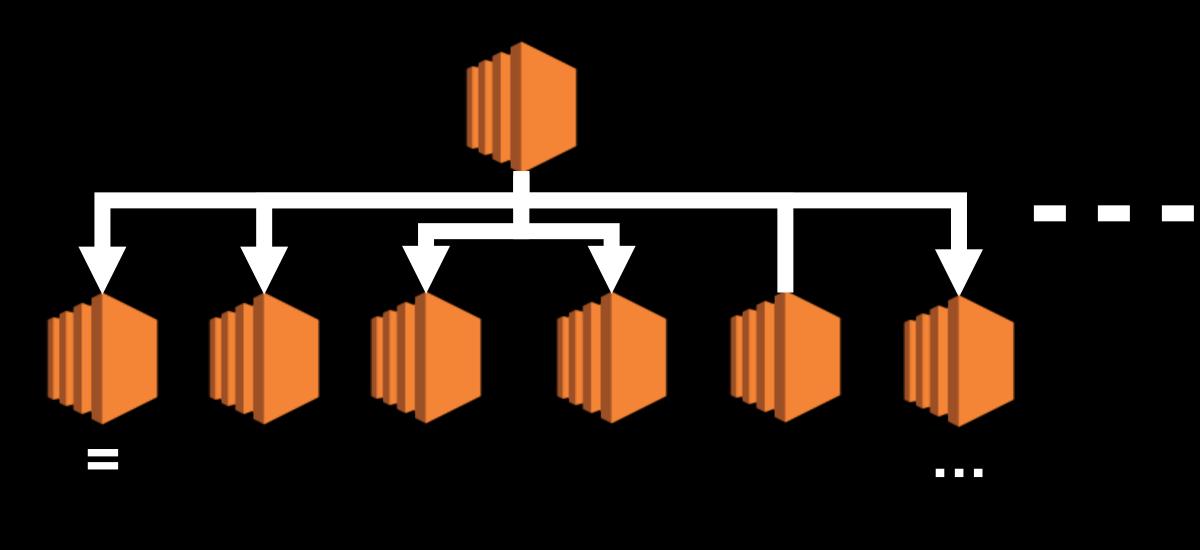
**Events** 



# ML pipeline: Spark ML/EMR/ElastiCache

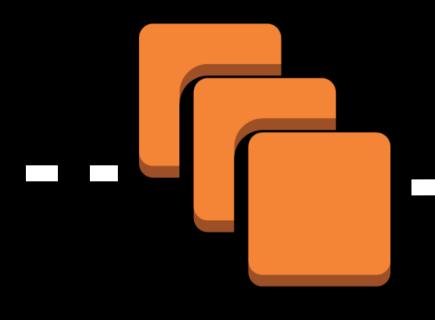




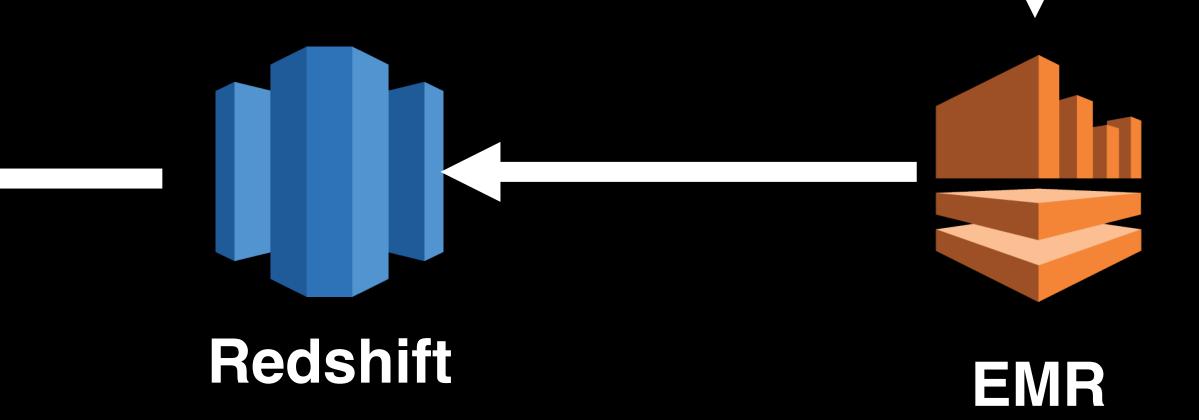




#### Report

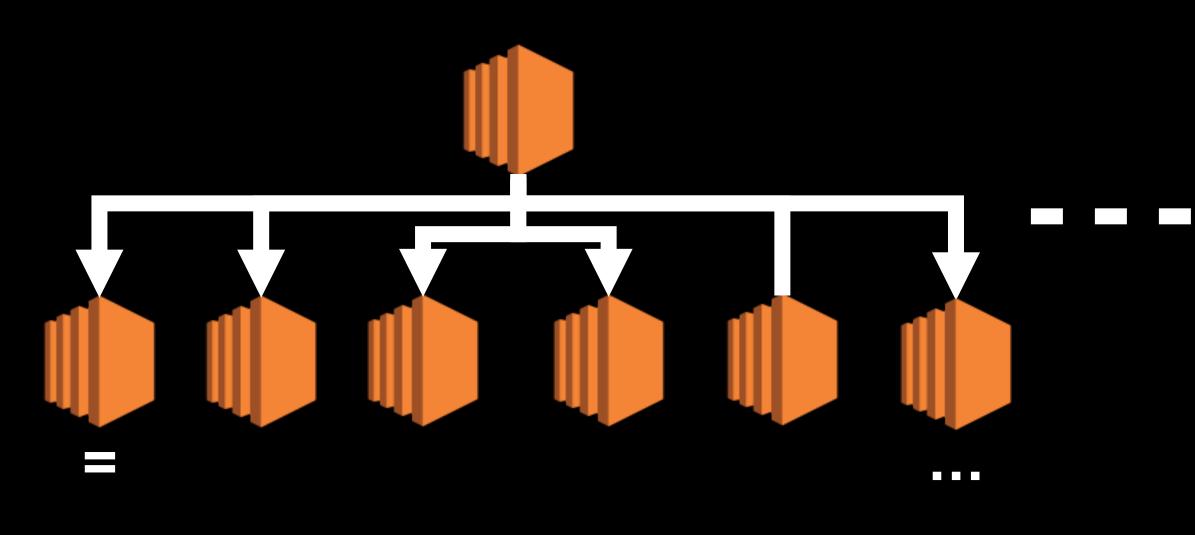


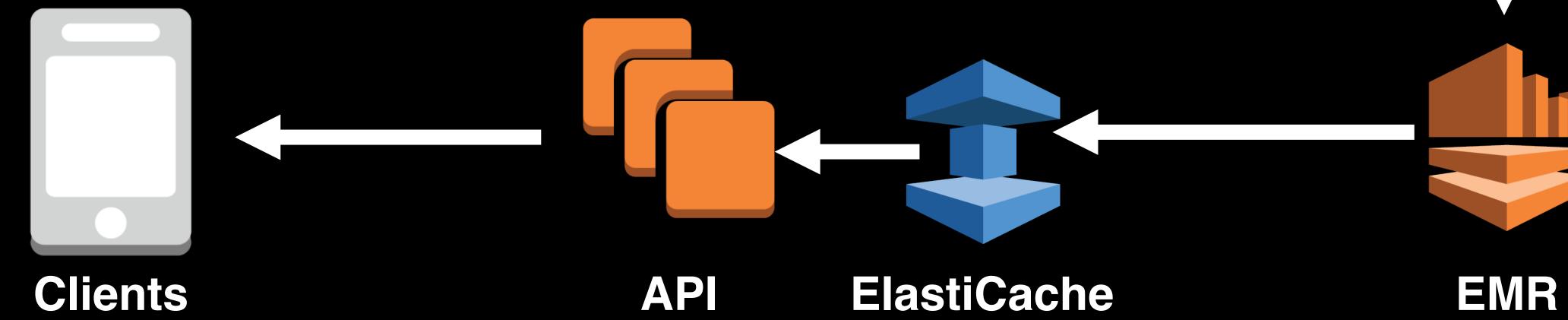
#### **Apache Flume**

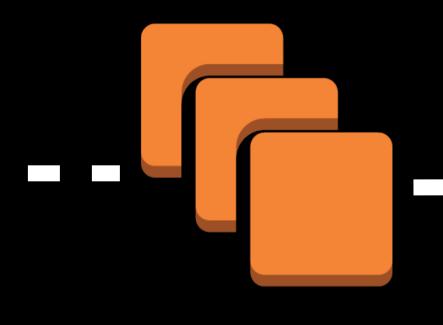


**Events** 









#### **Apache Flume**

**Events** 



#### Machines produce a ton of data

Hadoop management is time consuming, avoid it!

Event driven data architectures have many advantages



# Thank you!

