





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



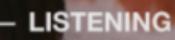
EyeEm

INDOORS

BRICK WALL

89% Aesthetic Score

WIRELESS TECHNOLOGY



CHEERFUL

SMARTPHONE

and the second sec

YOUNG WOMAN

IOLDING





Three Parts Research, Production, Data





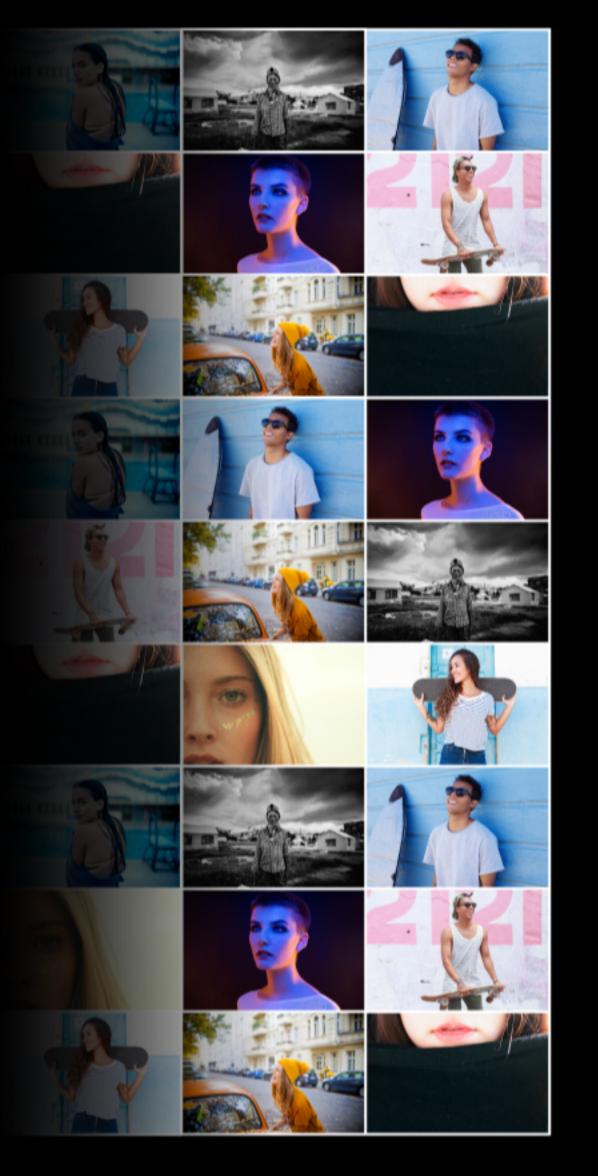
Beginning: Research

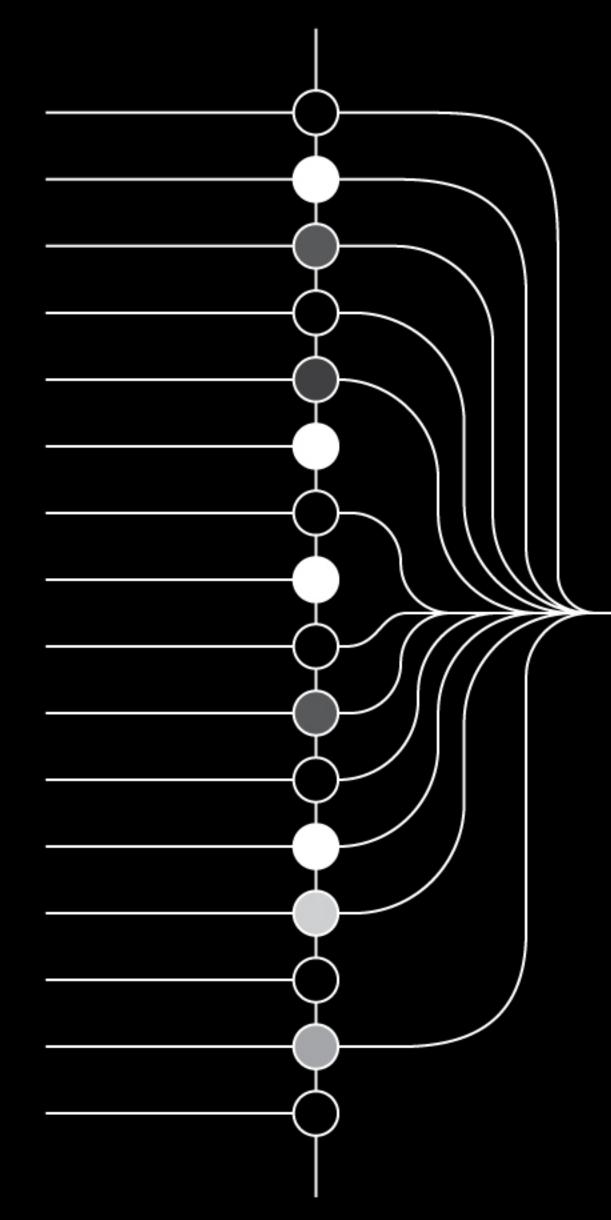
Where the models (and some trouble) come from

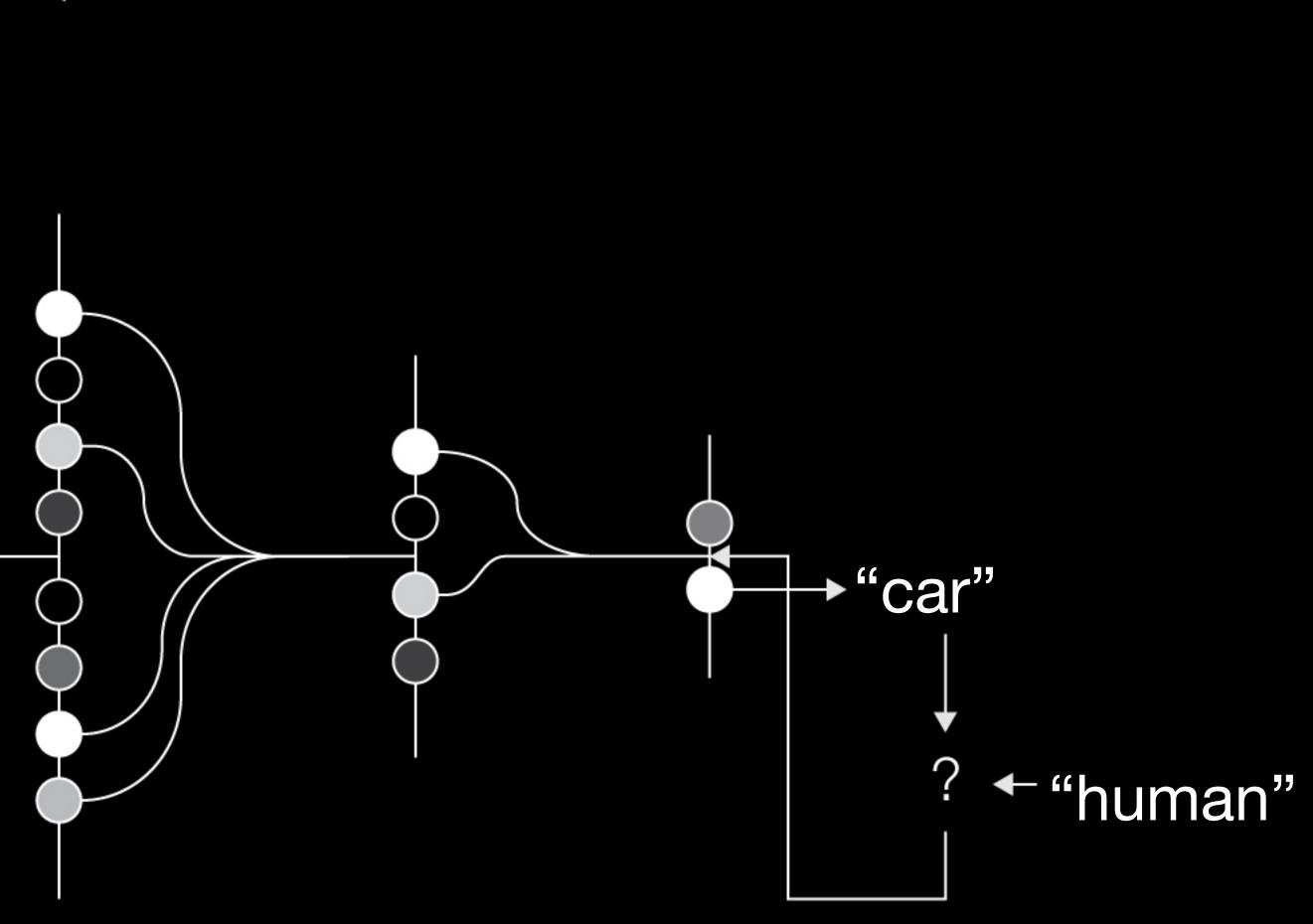


Training and inference

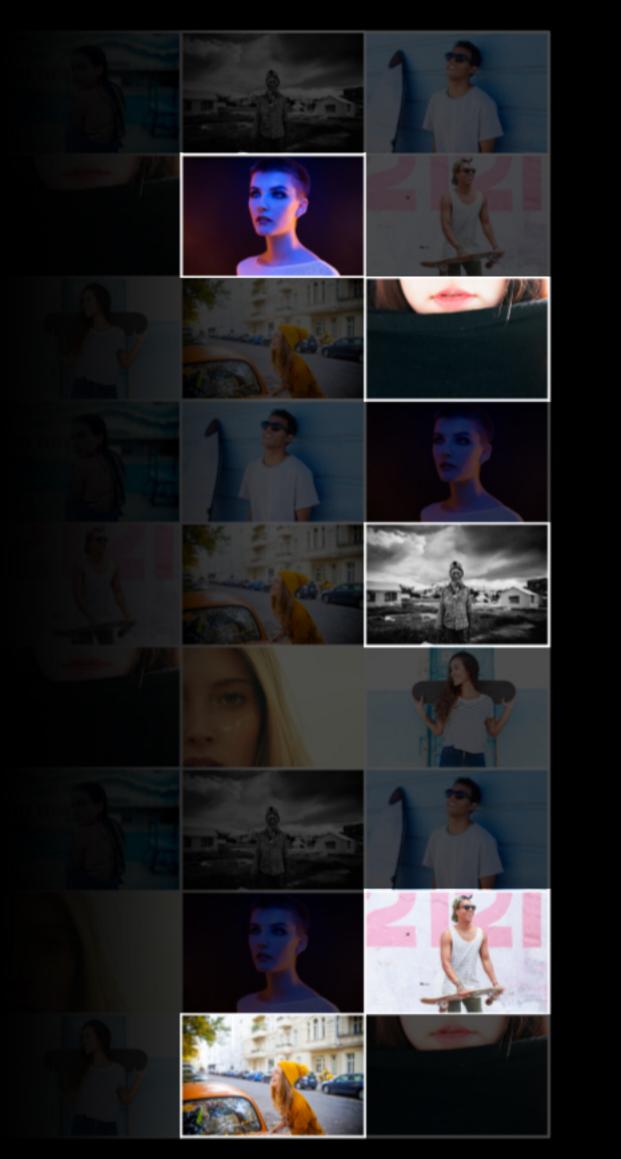


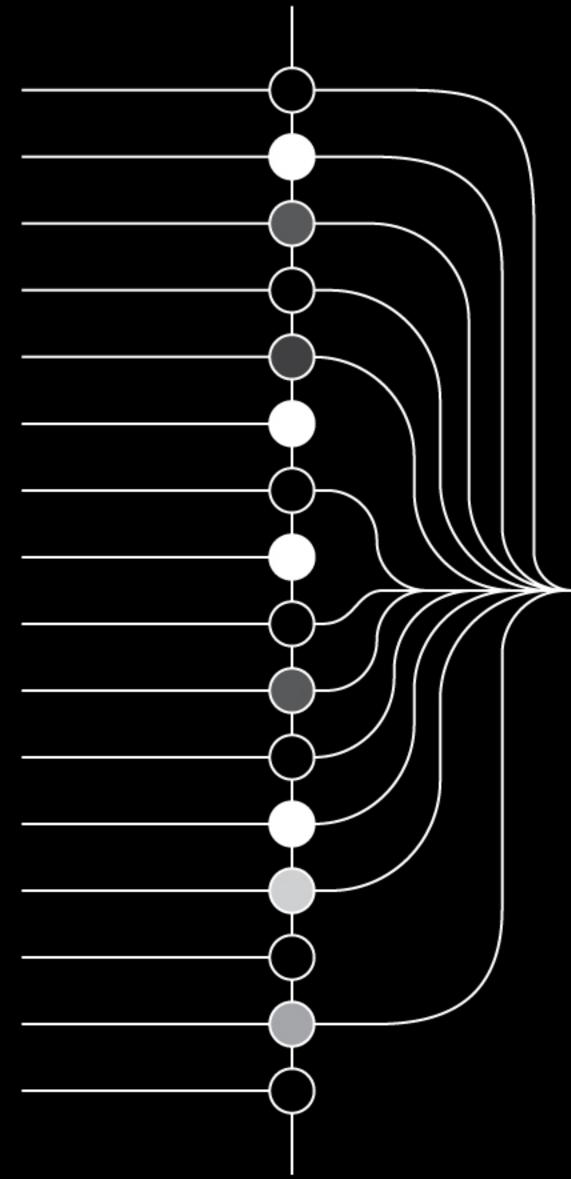


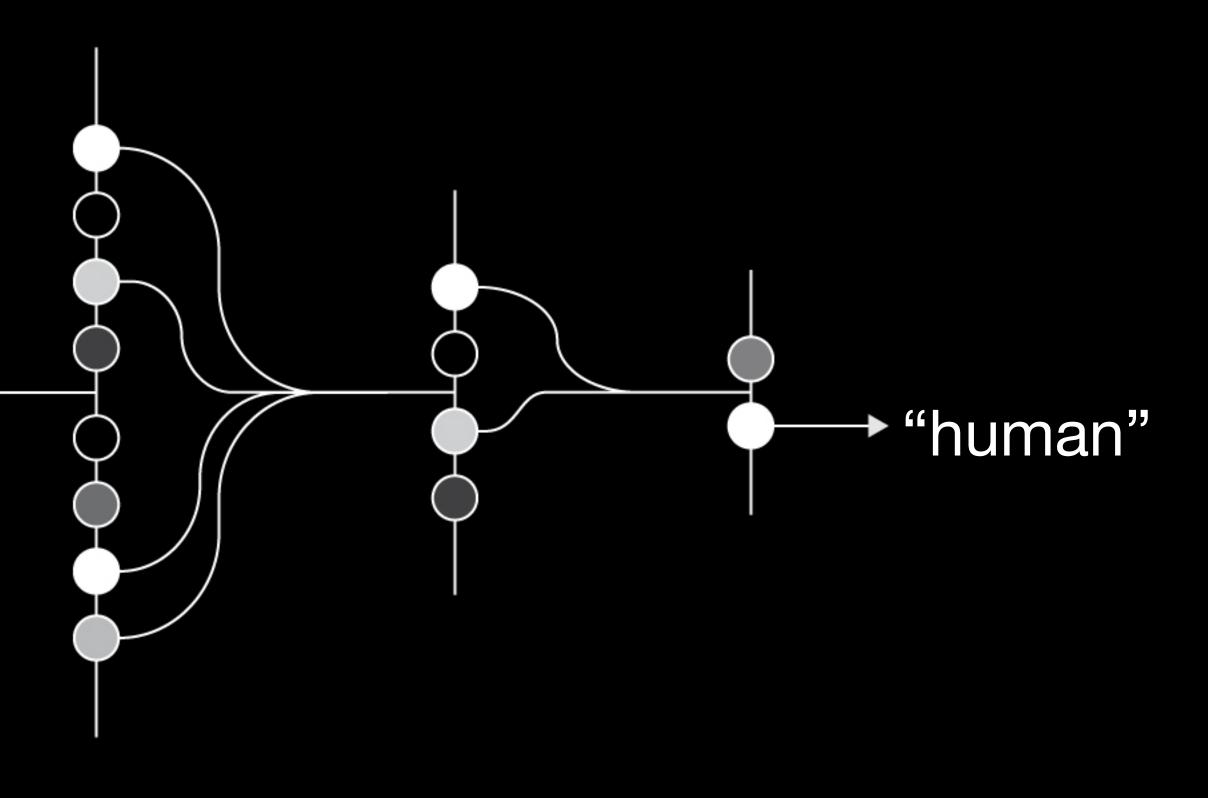














Deep learning and GPUs

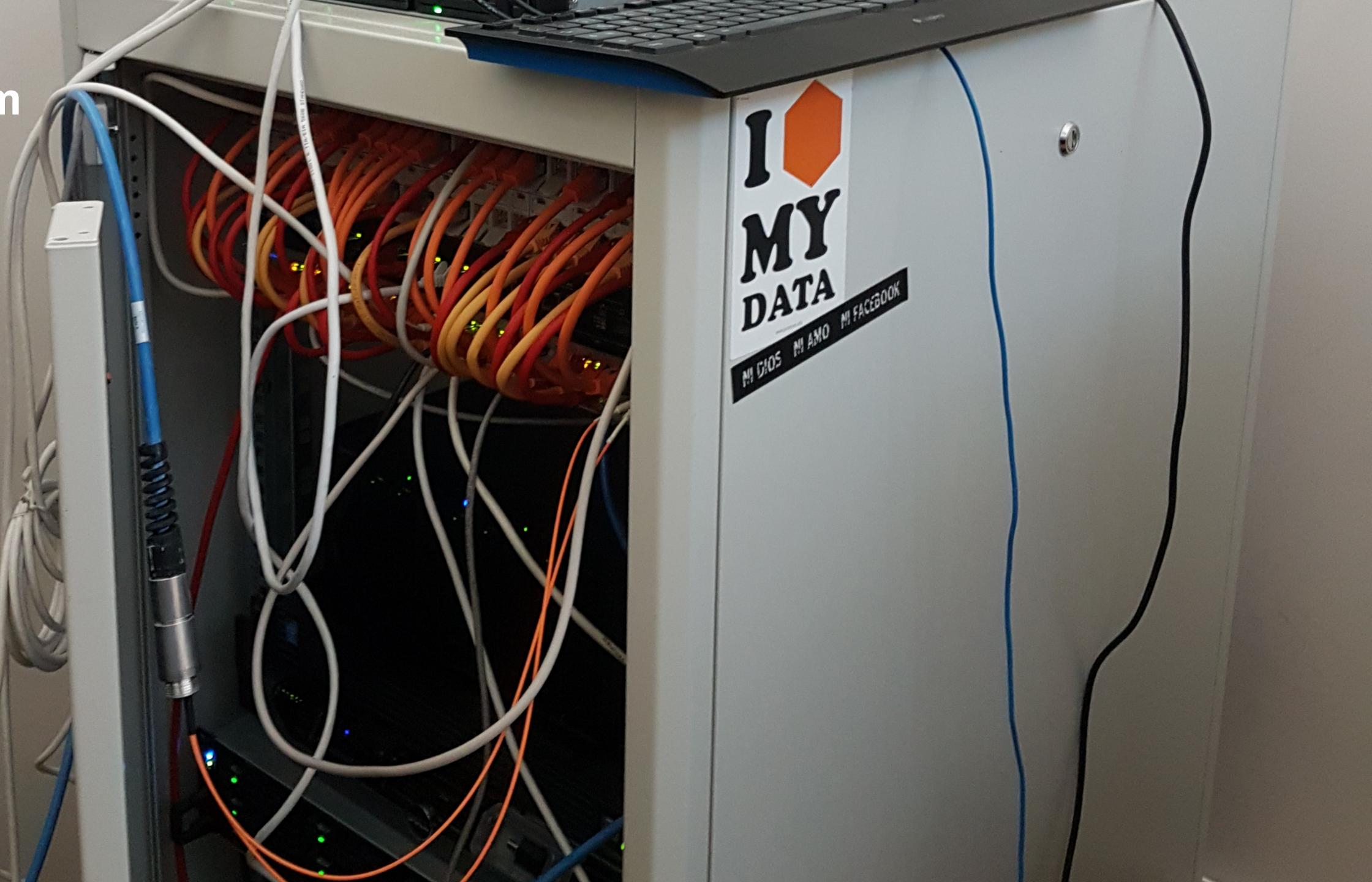
GPUs run deep learning training and inference very efficiently



Training with millions of photos needs GPUs

More data, more GPUs

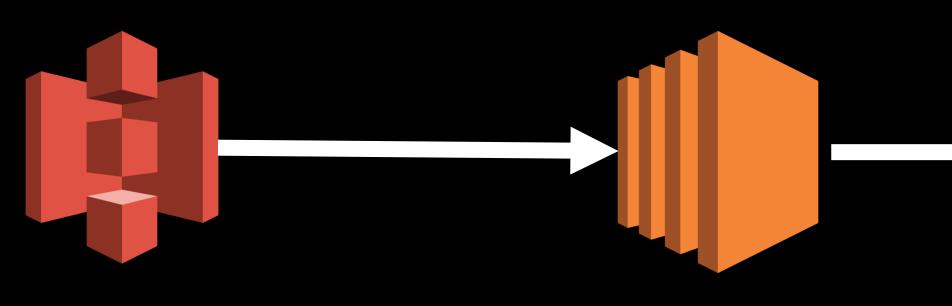
EyeEm





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



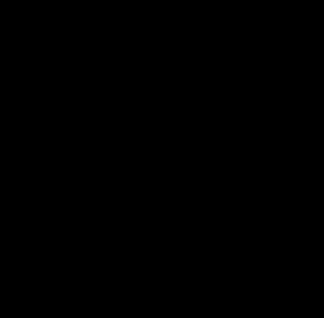
Problem Thumbnail Service

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 \$

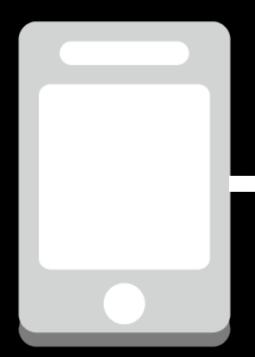


Middle: Production

Dealing with scale



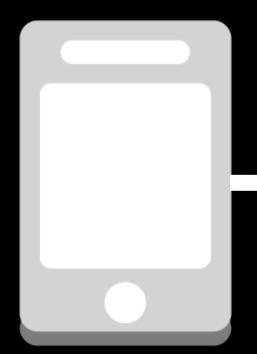














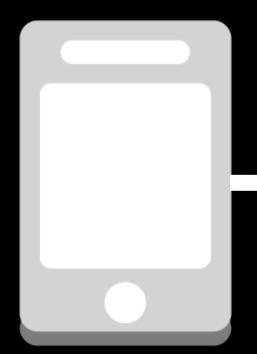


Search

Deep Learning Service

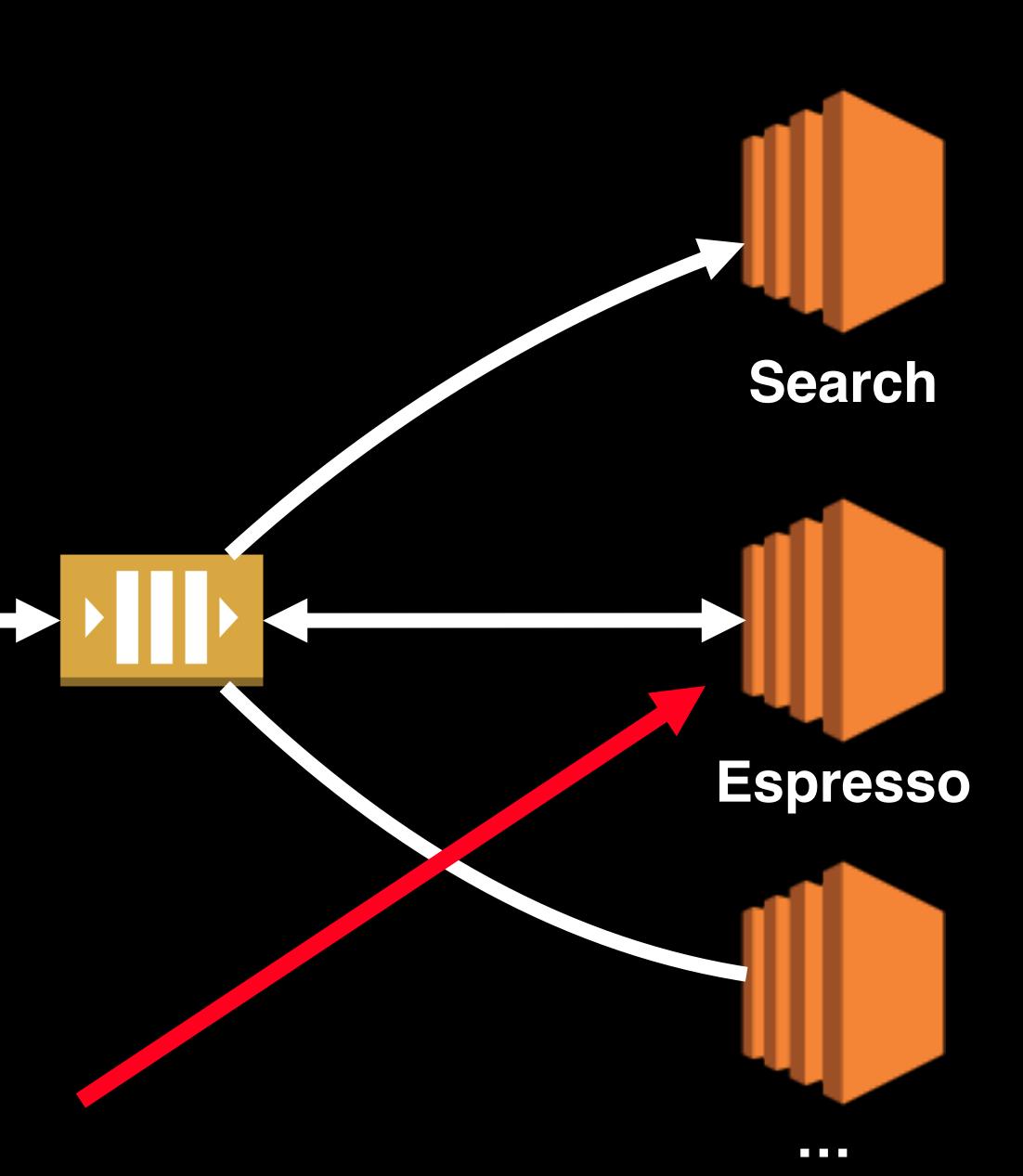




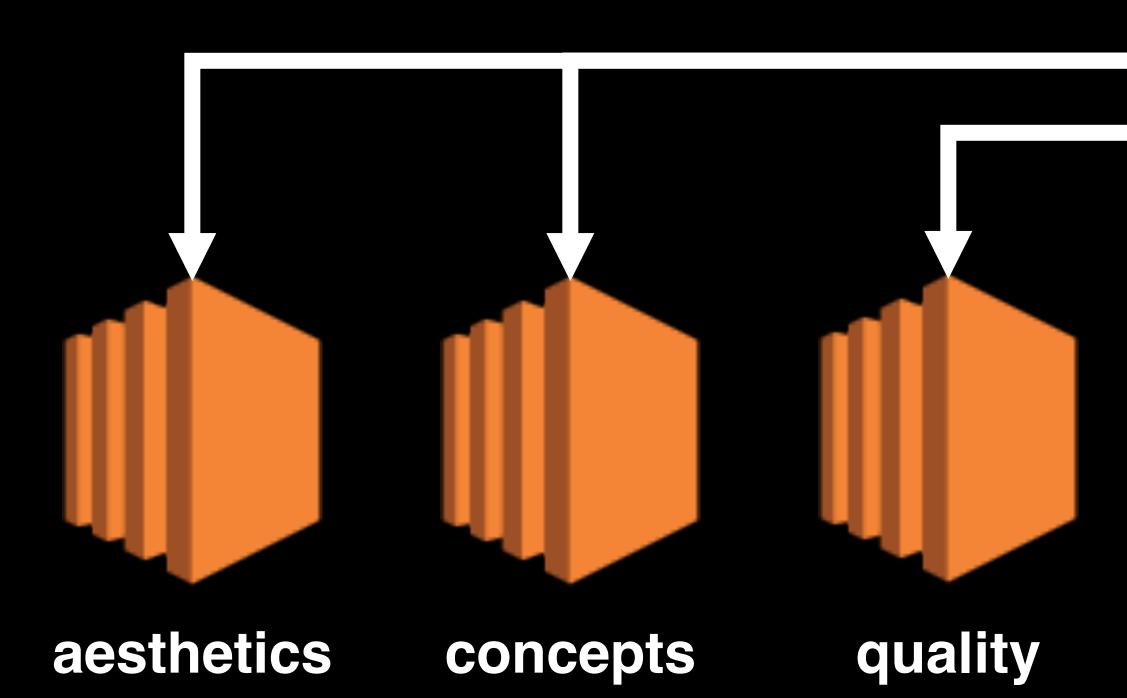












Espresso

G2 machines

captions

personalised aesthetics





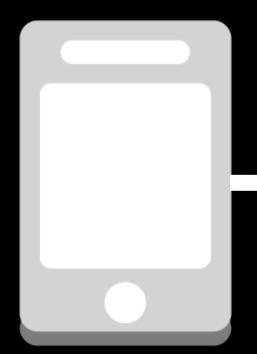


What we've automated







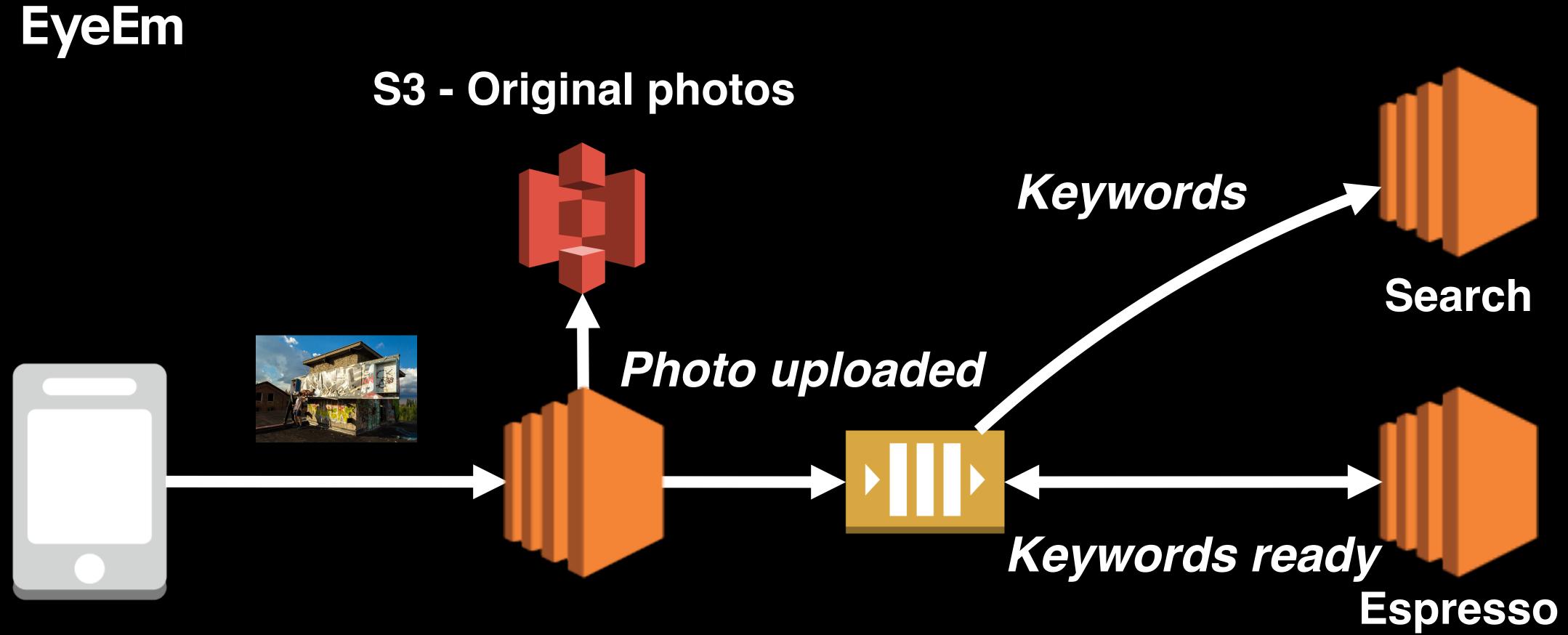












Edge

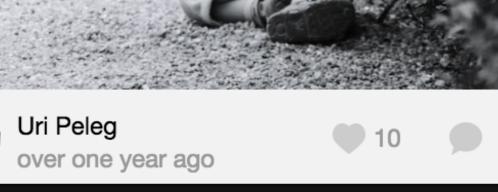
EyeEm	Following	happiness	Q	
Hide filters	- 15	3,612 results		
Market Photos on Eye	– Em Market			
Format Portrait Landscape Square Widescreen Panorama		Frik Nardini about one year	112	
Size Large Medium				

Need help?

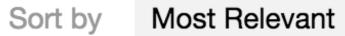
Small

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

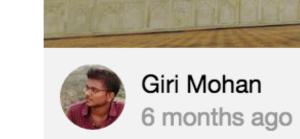
80



Market Dashboard



Madeline Yates 35

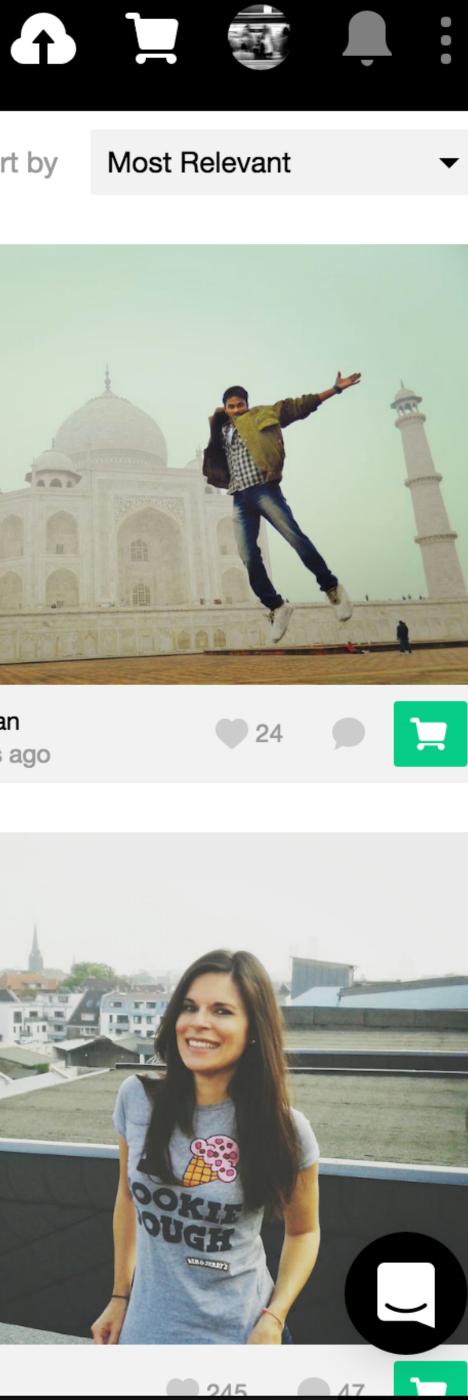








11 months ago

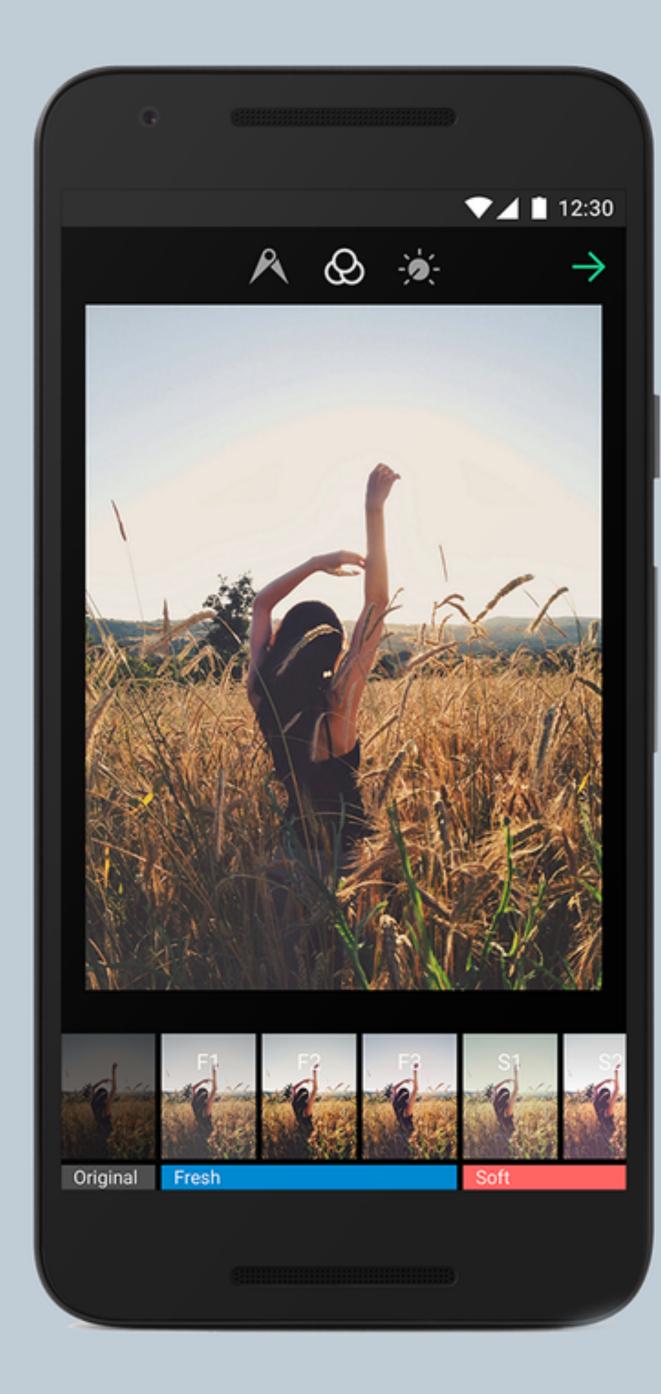


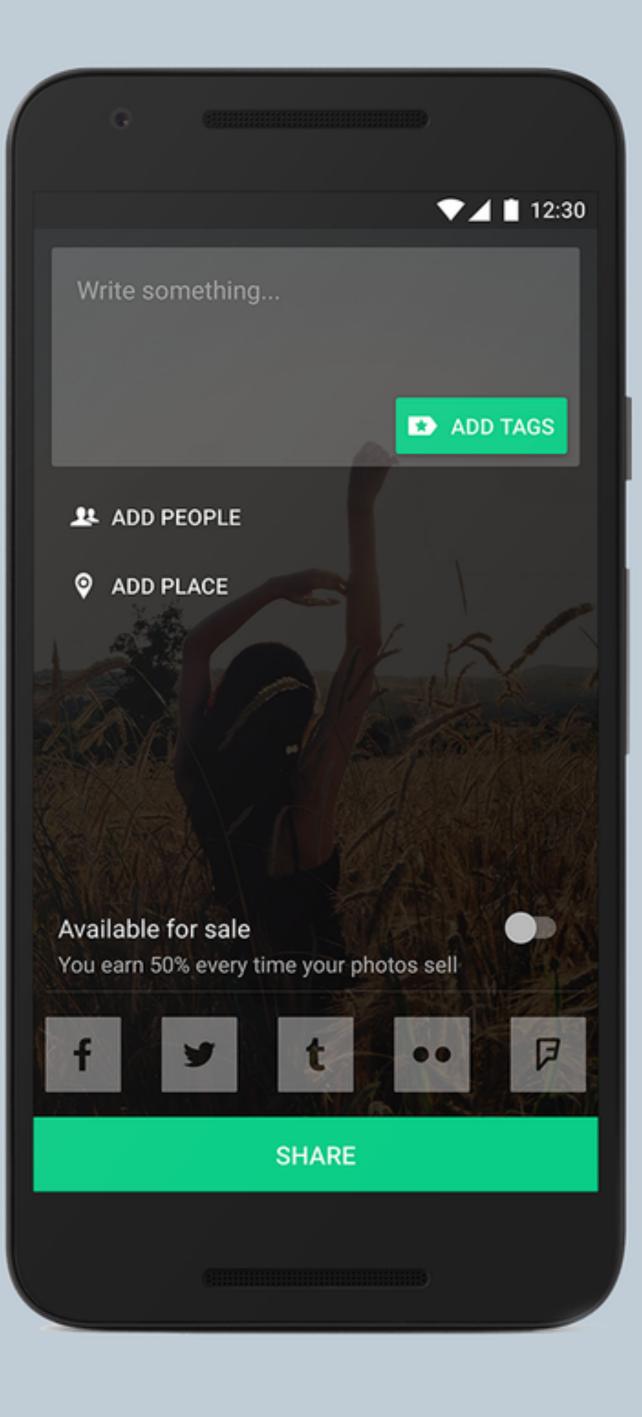


What we've automated

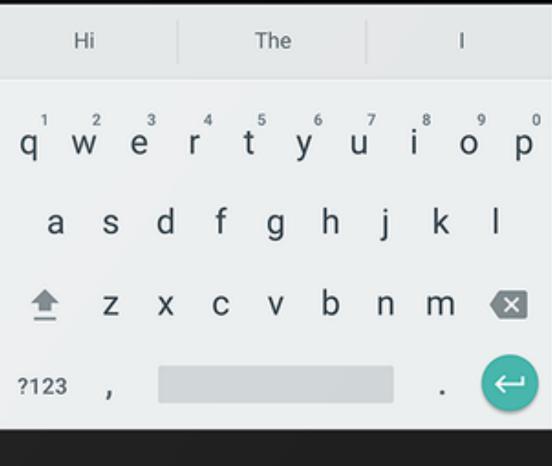
Auto-tagging







► 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



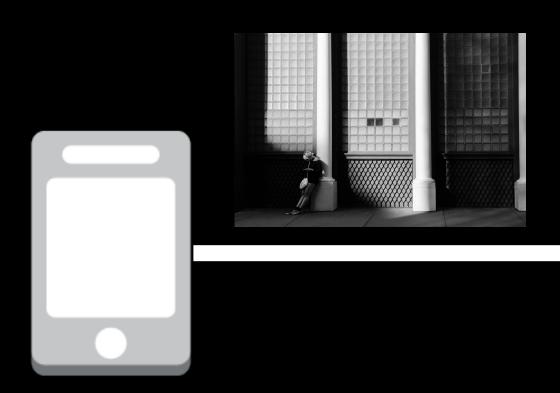
The traditional stock photography industry.

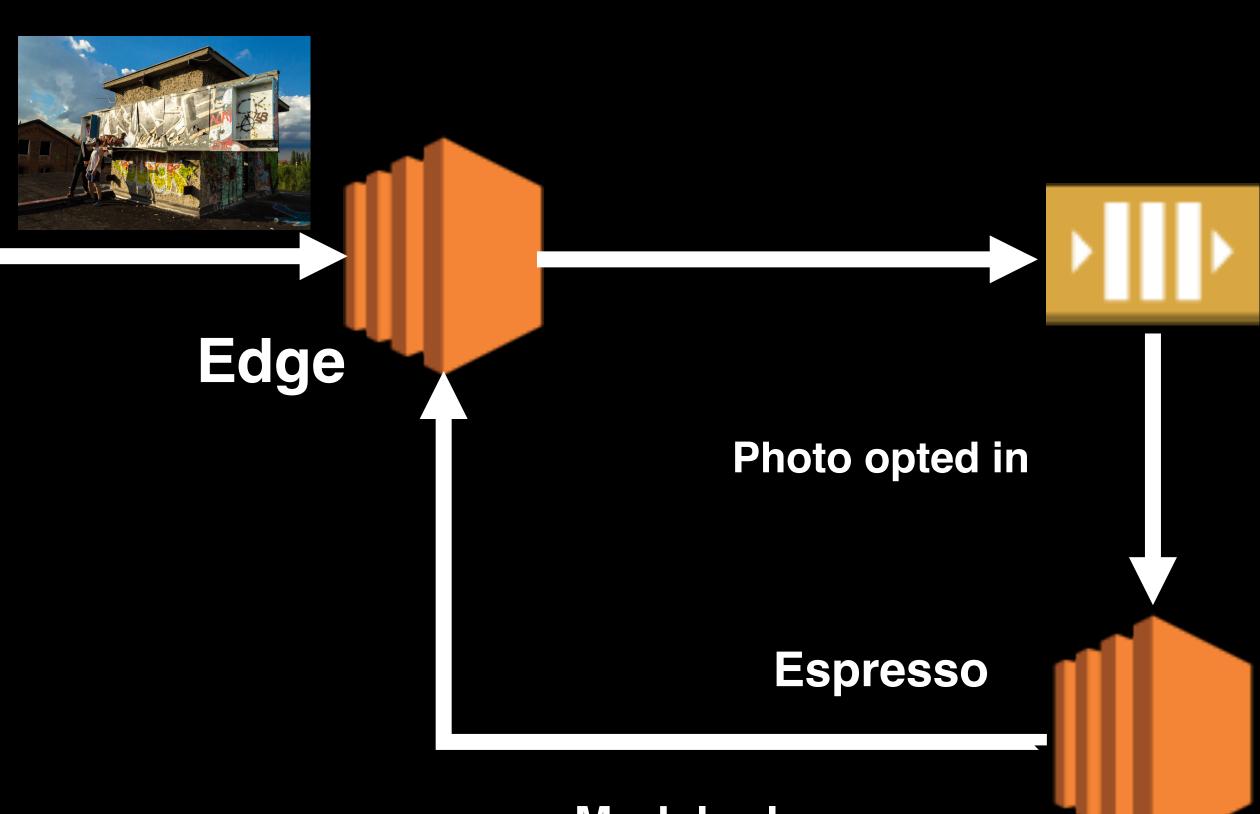




What we've automated Photo quality rating Model release detection

Photo reviewing Process





Model releases Quality **Keywords and captions**



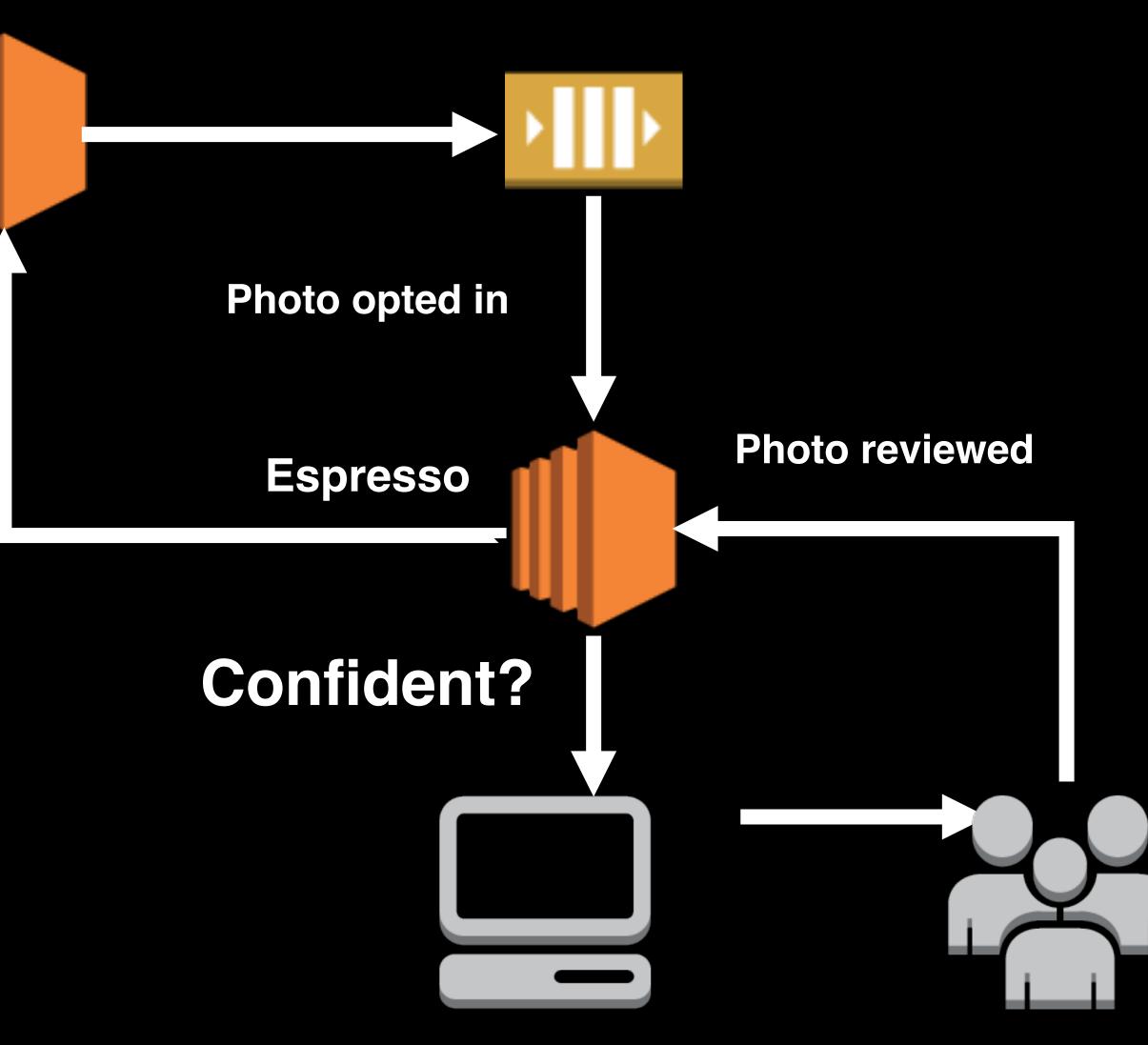
Woman photographing through smartphone

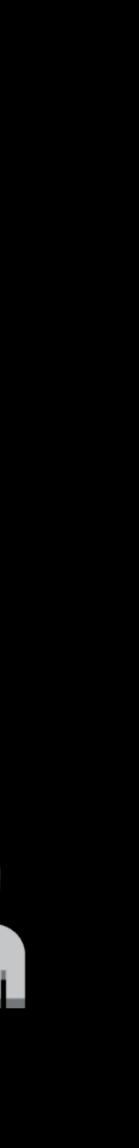




Model releases Quality Keywords and captions

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





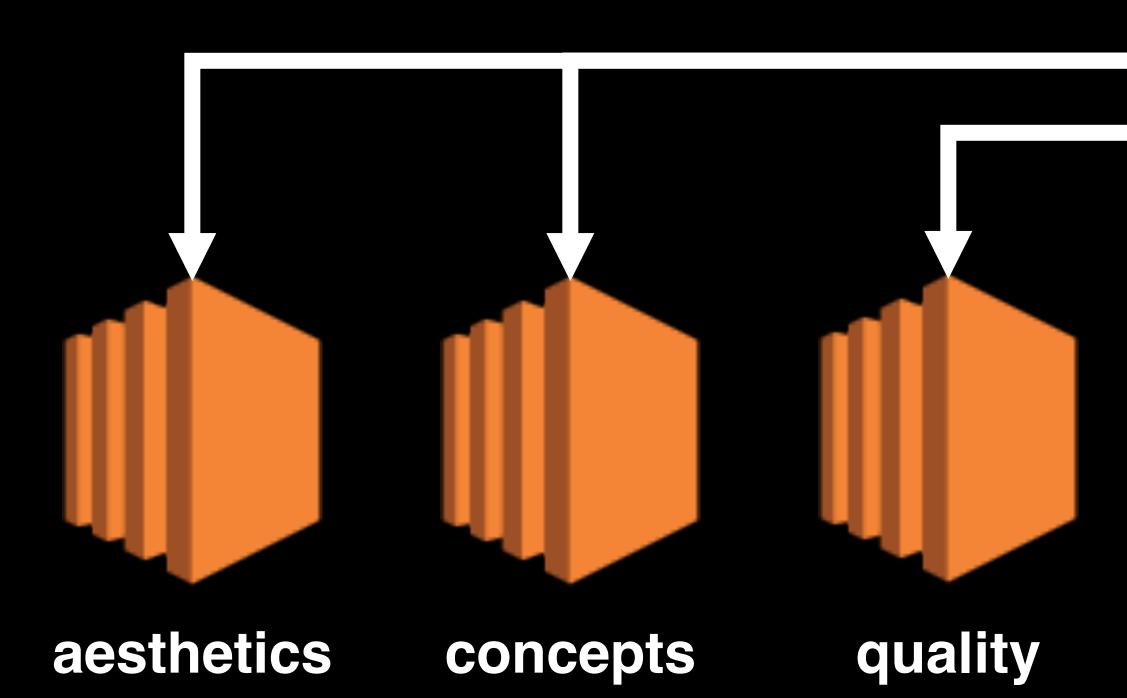
Everything runs



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

people", 0.99792659282684326], ["nature", 0.97948437929153442], ["sky", 0.9790 0015115737915], ["outdoors", 0.92300915718078613], ["day", 0.9213292598724365 2],["beach",0.91038638353347778],["scenics",0.88209074735641479],["lifest activity",0.80806028842926025],["sand",0.79079025983810425],["tranquility nature",0.69489455223083496],["landscape",0.64603018760681152],["tree",0. 48800194263458252],["mammal",0.31885015964508057]],"confidence":0.8<u>376338</u>

Concepts/Confidence



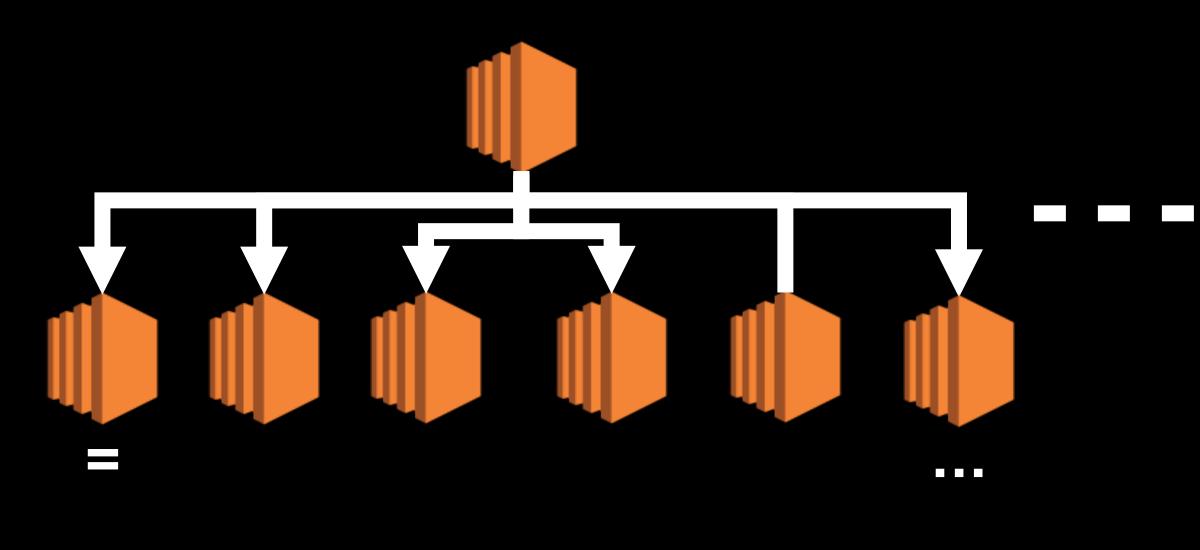
- {"photoId":"93900842","captions":[{"caption":"STREAM FLOWING THROUGH EYEREKS","probability":0.656147889422701,"wordsProbability"Capteons_
 - 0.47248415835501062],["FLOWING",-0.29332975536168115],["THROUGH",-
 - 0.59238967310011115], ["ROCKS", -1.1257078392955155]]}, {"caption": "VIEW OF STREAM FLOWING THROUGH
 - ROCKS", "probability":0.5243615522471853, "wordsProbability":[["VIEW",-
 - 1.8661552402619594], ["OF", -0.00410881194664866661], ["STREAM", -
 - 0.087614953179970007],["FLOWING",-1.3746558803077191],["THROUGH",-
 - 0.31274626783455156],["ROCKS",-1.5081047112344315]]},{"caption":"VIEW OF
 - STREAM IN
 - FOREST", "probability":0.4874233842518491, "wordsProbability":[["VIEW",-
 - 1.8661552402619594], ["OF", -0.00410881194664866661], ["STREAM", -
 - 0.087614953179970007], ["IN", -1.3630659191281458], ["FOREST", -
 - 1.690672657028897]]}, {"caption":"VIEW OF STREAM IN
 - WATER", "probability":0.43154049087804885, "wordsProbability":[["VIEW",-
 - 1.8661552402619594], ["OF", -0.00410881194664866661], ["STREAM", -
 - 0.087614953179970007], ["IN", -1.3630659191281458], ["WATER", -
 - 2.380402292097590211}1





Billions of events

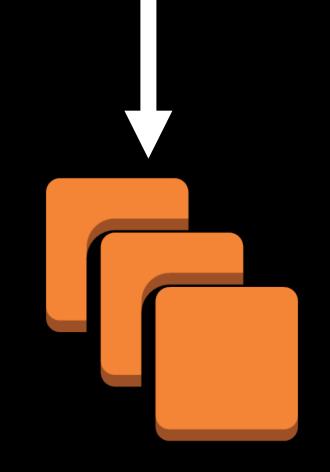






Report





HDFS/Spark



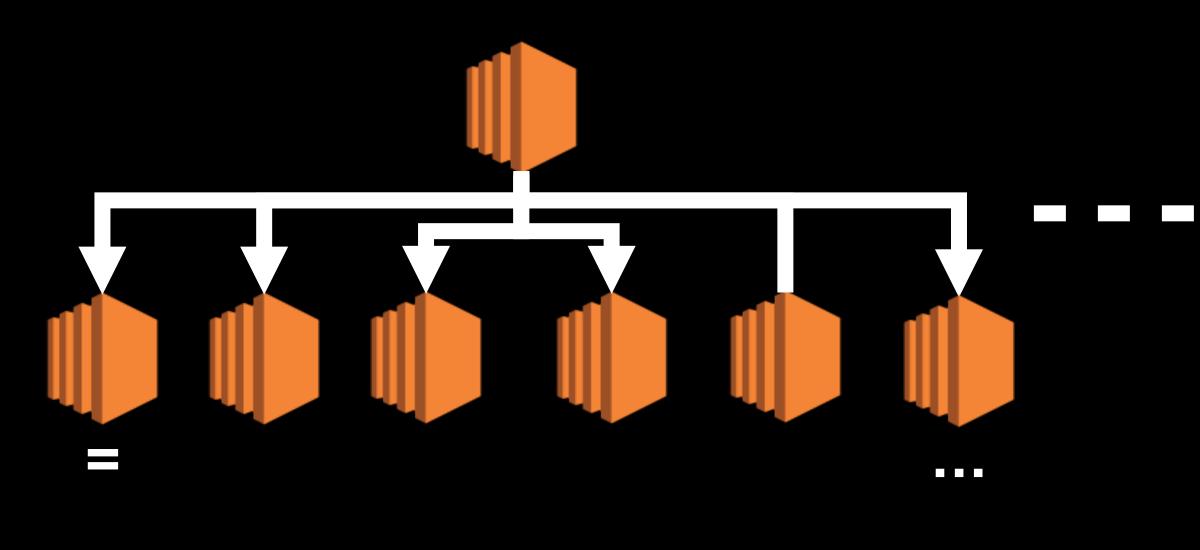


Our move to EMR



Analytics pipeline: Spark/EMR/Redshift

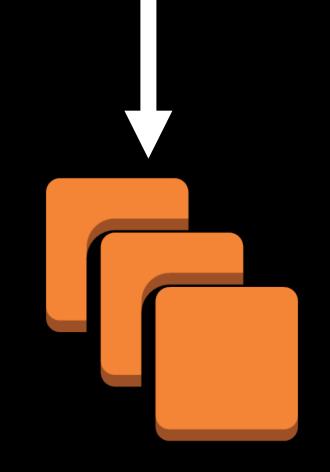






Report

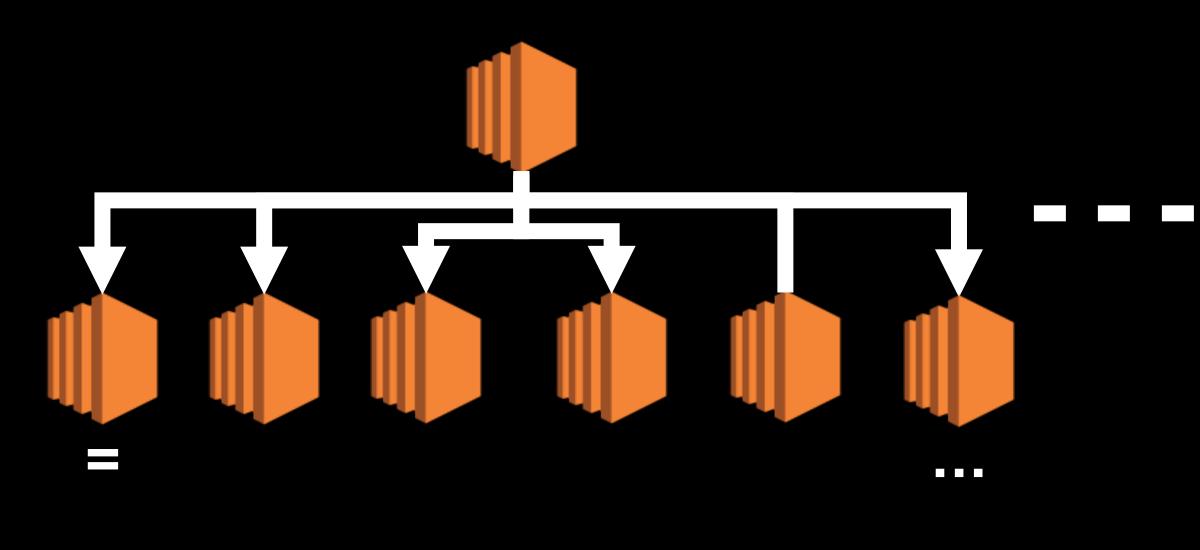




HDFS/Spark

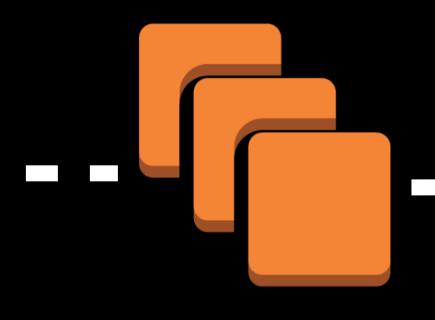




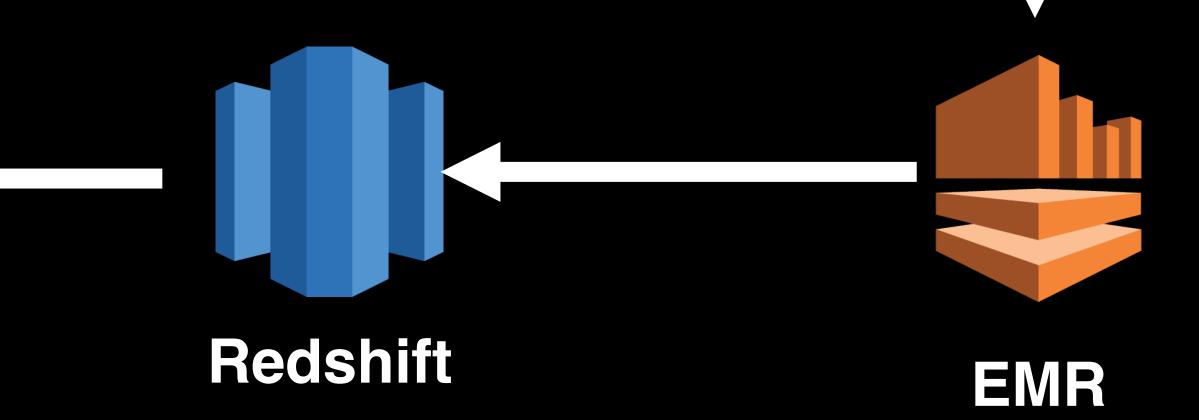




Report



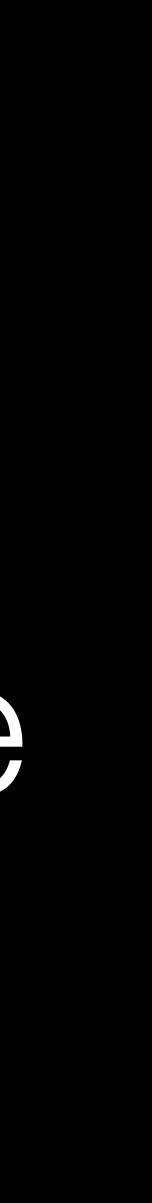
Apache Flume



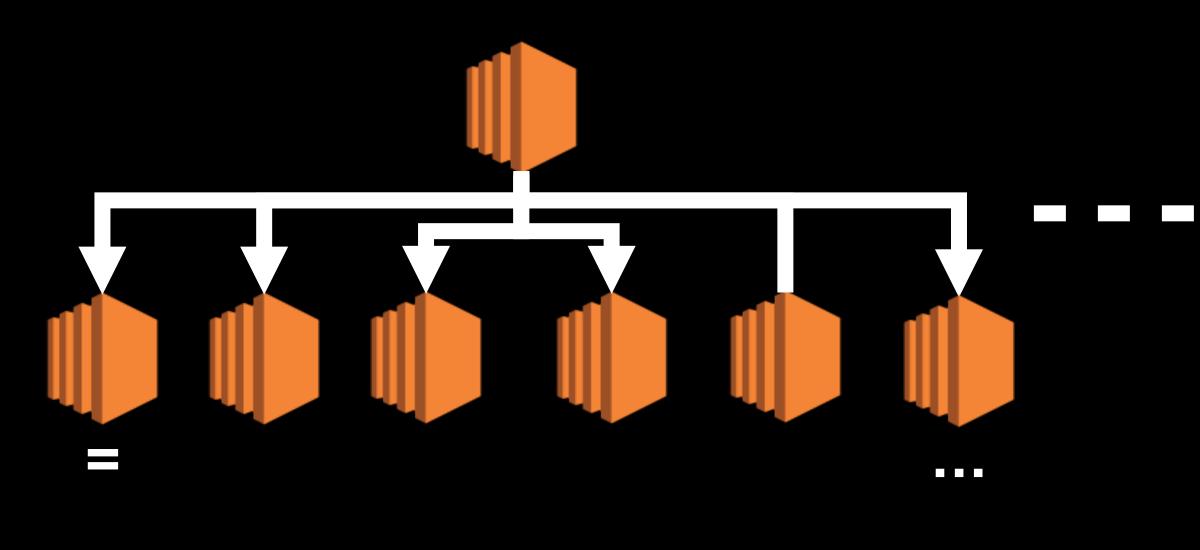
Events



ML pipeline: Spark ML/EMR/ElastiCache

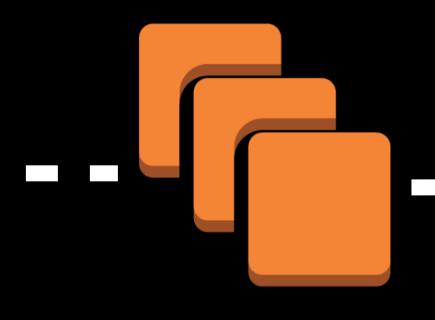




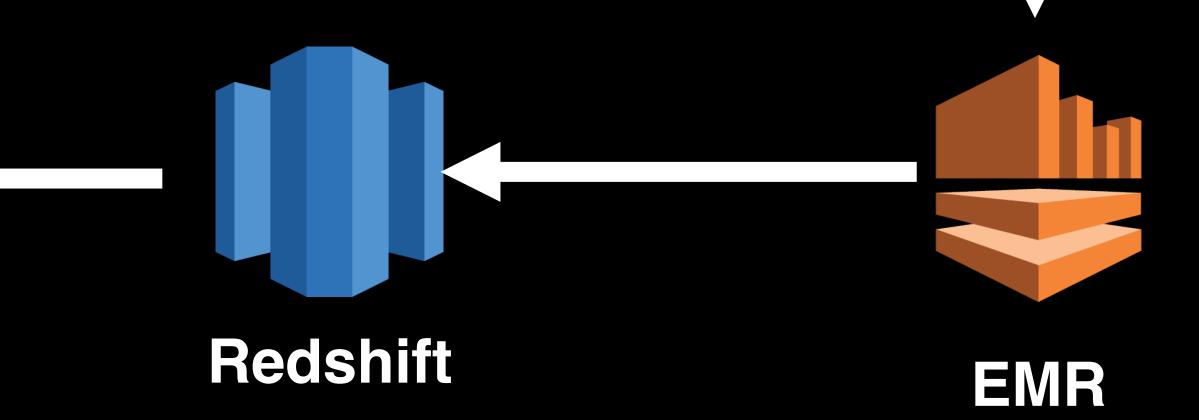




Report

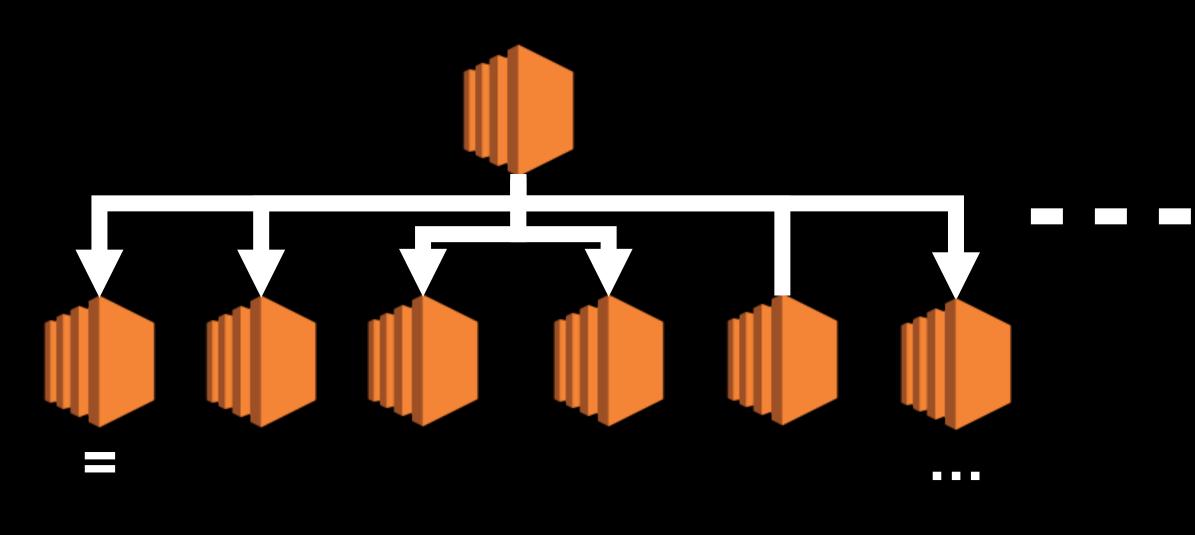


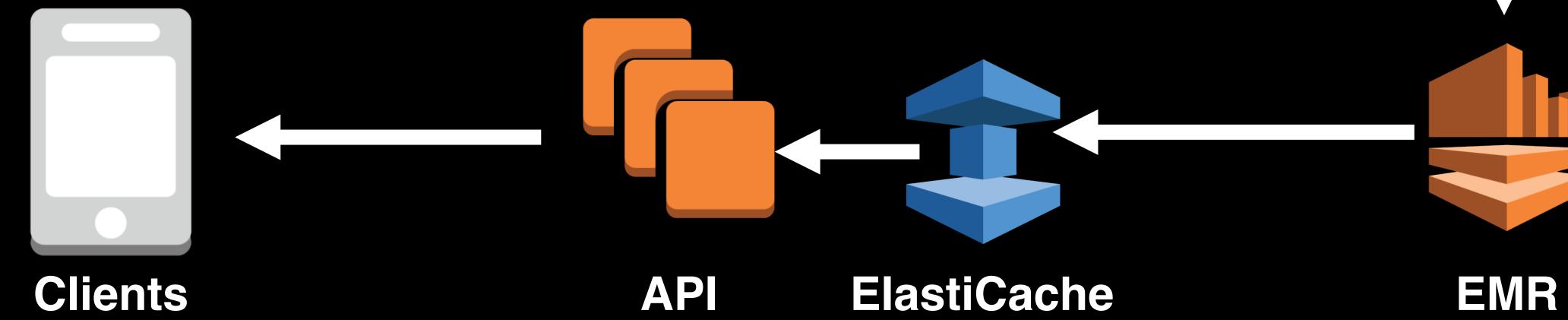
Apache Flume

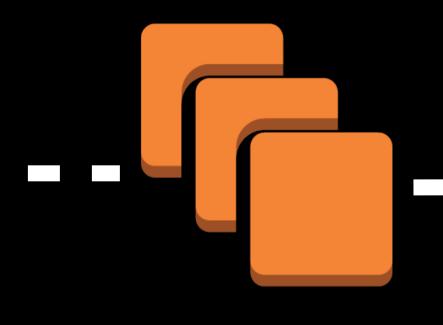


Events









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!

