

Using Machine Learning Power of Google's Cloud Vision API to

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EXECUTIVE SUMMARY

Introduction

In this research we used the machine learning technology of Google's Cloud Vision API for the first time ever to test ad effectiveness. We tested over 9.000 online ads from over 50 product categories. We ran these ads through the Vision API to find out what ad characteritics drive the most success in terms of CTR. As far as we know, this is the first ad effectiveness study using Machine Learning as the basis of the methodology in the Netherlands.

Methodology

To collect all ads and their performance metrics from our ad servers we used the DCM/DFA Reporting and Trafficking API and Java client libraries. We loaded all creatives for which performance metrics were still available, and stored them specified per day. We uploaded all images into Google Cloud Vision API requesting Label, Text, Landmark, Logo Detection, Face Detection and Dominant Colors, and stored the results for analysis. Google provided funding for this research.

Most important findings

- **Size** does matter: the bigger ad format the better the results
- □ Horizontal is the best shape
- The best colors to use are blue and black
- □ A logo adds value and increases CTR
- □ The suggested position of the logo is bottom-left
- Either few or a lot of **text** increases the chance on a high CTR
- A call to action works: 'win' or 'bekijk' work best
- Multiple CTA's are contra productive: it scores lower than one CTA

AD EFFECTIVENESS USING MACHINE LEARNING TECHNIQUES

Google Images Search

When you use Google Search images, and you search for 'blue t-shirts', you will get images with blue t-shirts in it, even though they are not specifically tagged with 'blue' and 't-shirt'. The technology used by Google is trained to recognize visual characteristics of images, using machine learning. The machine recognizes color, logos, texts, fonts, faces, animals, et cetera. It also recognizes context and can, for instance, associate pink with 'female' in a fitting context.

Google Cloud Vision API

We used this technology to test the effectiveness and characteristics of thousands of online display and animated ads. And we studied the impact of these characteristics on the click through rate (CTR) that the ads have had. As far as we know, this is the first ad effectiveness study using Machine Learning as the basis of the methodology in the Netherlands.





METHODOLOGY

Getting all images from DCM

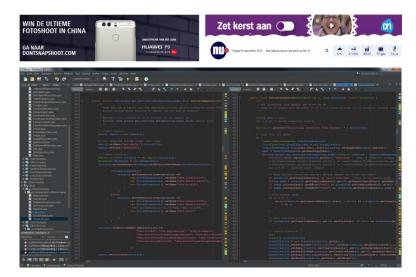
To collect all ads and their performance metrics from our ad servers we used the DCM/DFA Reporting and Trafficking API and Java client libraries. We loaded all creatives for which performance metrics were still available, and stored them specified per day. For animated ads, the URL's where reconstructed and then each of the ads was automatically opened for 30 seconds using PhantomJs or SlimerJs. During these 30 seconds, screen captures where taken 10 times each second, giving us up to 300 frames for each ad.

Getting key frames from animations

We created a python script that analyzes animations based on the frame to frame changes to select important frames from the 300 we collected before. From this we also acquired properties of the animation, like animation length, the number of loops and the type of animation (continuous or presentation). The script also calculated the number of unique colors in each frame, whereas a low number of unique colors indicates a large amount of text or drawings, and a high number of unique colors indicates photography.

Running Cloud Vision API

Finally we uploaded all images into Google Cloud Vision API requesting Label, Text, Landmark, Logo Detection, Face Detection and Dominant Colors, and stored the results for analysis.





CHARACTERISTICS

What did we test?

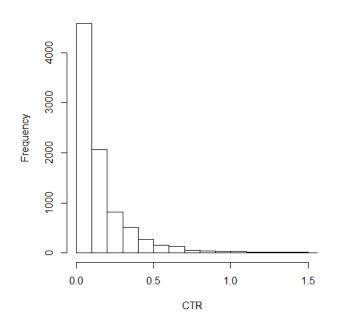
We tested over 9.000 ads and ran them through Cloud Vision API. Only such ads were included that had at least one click and more then 5.000 impressions. For video formats we created a script that broke the video down into frames.

The machine came back with an output of characteristics such as colors (RGB), size, fonts, logo, logo positions, text, text in capitals, client category, call to action (CTA).

The average CTR of all ads was 0,28%.

	CTR (%)	
Min	0.000232	
Мах	19.73185	
Mean	0.284659	
Median	0.097299	

Click Through Rate



LABELING ASSOCIATIONS & RECOGNIZING OBJECTS

From object recognition to facial expressions

It is definitely interesting to see how "the machine" works and what associations it makes. It not only measures the number of letters, pixels, RGB colors, fonts and logos, but also recognizes animals, famous buildings like the Eiffel Tower, objects like a sail boat or a car, and facial emotions of people like joy, sorrow and anger. Moreover, Cloud Vision API associates colors and visuals. An example from our analysis: An ad with a dominant pink color and a heart shape in it was not only labeled 'pink' and 'heart' but also as 'female'.

Most of the time the application is right, like when it recognizes that the JBL ad on the top-right displays clothing, a swimmer, etc. However, sometimes we see some strange associations, like with the organ donor ad on the bottom-right. Probably because of the limited and zoomed image of the bodies, Cloud Vision API was not able to recognize the human bodies in it.



Clothing	Swimmer	Muscle	Sports	Water sport
0,902	0,814	0,725	0,649	0,575

Wa kunnen i	nt zou jij missen?			WERKING VAN JE ORGANEN?
Clothing	Wallet	headgear	Personal	protective equipment
0,981	0,596	0,555	0,509	

LABELING ASSOCIATIONS & RECOGNIZING OBJECTS

The limits of Vision API

We found that the Cloud Vision API not always understands all elements in the ads. Many times it labeled the ads just as 'advertising', which is correct C.

It didn't recognize 'hearing device' and 'lip stick'. Although the hearing device was not recognized as such, the Vision API understood that it is a small appliance. The rest of the associations were wrong.

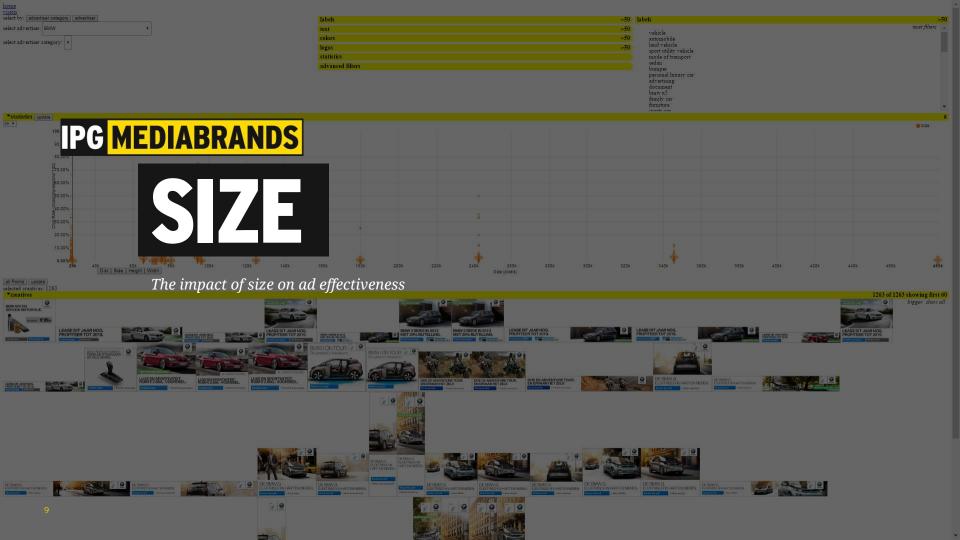
The lipstick surprised us to be honest. It is such an ordinary shape. Vision API is correct in understanding 'Lip' but is wrong on the other associations. Text is correct, but 'Major Appliance' and 'Gadget' isn't.



ASSOCIATION	LIKELIHOOD
Small appliance	0,553
Home appliance	0,550
Plumbing fixture	0,530
Energy	0,502



Text	0,950
Major appliance	0,789
Feature phone	0,777
Gadget	0,743
Lip	0,708

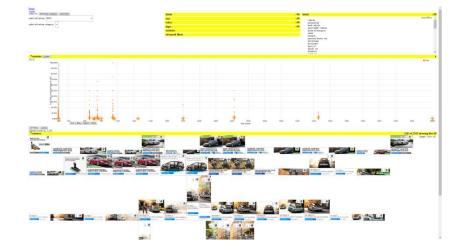


THE IMPACT OF AD SIZE

The impact of size: the bigger the better & horizontal is good

Most of the ads we tested were in the same size range (50.000 and 100.000 pixels), with an average CTR of 0,1222. Larger ads tend to have a higher average CTR. The 300.000-350.000 pixel-sized ads have an average CTR of 3,36.

Moreover, the shape of the ad has serious impact on its CTR. Horizontal ads are more effective (average CTR of 0,3275) than vertical and squared ones, though we have to be mindful that there were only 23 squared ads in our data set.





	Number of occurrences	CTR (%)
Vertical	2807	0.1917
Horizontal	6180	0.3275
Squared	23	0.1065

IPG MEDIABRANDS



The impact of color on ad effectiveness

THE IMPACT OF DOMINANT AD COLOUR

The impact of color: blue and black

We tested the impact of the dominant colour of the ad. The dominant colour of an ad with multiple frames, is the most common color in all the frames in our calculations. On average, 14.044 colours (RGB values) are used in the ads.

Highest CTR

The colours with the highest CTR are blue and black. Red, green and white perform less with small differences between each other.

Color	#	CTR (%)
red	3158	0,2538
blue	2003	0,3539
green	2152	0,2418
black	426	0,3698
white	1270	0,2964

Mean colors			
Min	2		
Max	174623		
Mean	14044		
Median	8676		

EXAMPLES DOMINANT COLOURS



Dominant colour=blue

13 Title of presentation



Dominant colour=red

Dominant colour=white

Coca:Cola

#ShareaCoke.

Typ hier je naam Check →

Ontdek of jij er bij bent!



Dominant colour=green



Dominant colour=black



SELFIDEO CLIP HEI KAAANIJE PAPPIE IN DE SCOOL

> KLIK HIER



The impact of a logo on ad effectiveness

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-05

TED BAKE

THE IMPACT OF A LOGO AND ITS POSITION

A logo delivers extra CTR

66% of the ads contain a logo and these ads perform better than the ads without a logo. Some ads (9%) even have multiple logo's, which did not add any performance but was even detrimental to CTR. Moreover, the use of multiple logos can confuse Cloud Vision API.

Positioning the logo

The lower left corner is the best position in our test, with an CTR of 0,53. Second best is to place the logo at the upper left corner of the ad.

#	CTR (%)
5970	0.3018
3040	0.2509
ш	
#	CTR (%)
798	0.2653
8212	0.2865
	5970 3040 # 798

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CTR (%)		Position
0.43563	725	Upper left
0.16233	276	Upper right
0.40654	1255	Lower right
0.53244	828	Lower left
0.24234	424	Above
0.21206	381	Left
0.15284	848	Under
0.10802	329	Right
0.32050	604	middle
	 (%) 0.43563 0.16233 0.40654 0.53244 0.24234 0.21206 0.15284 0.10802 	(%) 725 0.43563 276 0.16233 1255 0.40654 828 0.53244 424 0.24234 381 0.21206 848 0.15284 329 0.10802









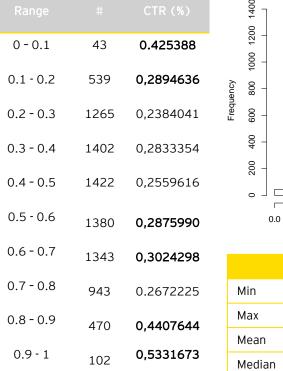
THE IMPACT OF THE USE OF TEXT

No clear impact of text

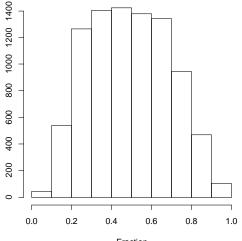
Text seems to have a positive impact on the CTR of online ads. This works in two ways: no or almost no text results in a higher CTR, as well as a lot of text.

CAPS

We didn't find a clear impact of the use of CAPITALS.



fraction of no text in ad



Fraction

Fraction of no text in ad				
Min	0,04774			
Max	1			
Mean	0.49040			
Median	0.48650			



THE IMPACT OF A CALL TO ACTION

Will a call to action in an ad increase the CTR? We defined six variables to flag the existence of commonly occurring Calls to Action (CTAs):

- Click
- Download
- Win
- Ontdek
- Bekijk
- Meer (lees meer, meer weten, meer voordeel, meer info, meer deals)

CTR is highest if the ad contains 'win' (0,534) or 'bekijk' (0,4). We found some ads with up to 3 different CTAs, but did not find a higher CTR when multiple CTAs are used. Knowing that the average CTR for ads without CTA is 0,24 the positive impact of a CTA is clear.

	Klik	Download	Win	Ontdek	Bekijk	Meer	СТА*	>1
Perc	11%	3%	1%	14%	10%	11%	46%	5%
CTA (%)	0.373	0.369	0.534	0.292	0.418	0.229	0.337	0.278

* Sum of CTA's included

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THE EFFECTS OF FACIAL EXPRESSION



Although the Vision API is very capable of recognizing facial expressions, we didn't have enough data points (not enough ads with faces in it) to find clear results. Our hypothesis was that happy faces result in higher CTR's.

See below the correct interpretation of Vision API regarding facial expression of the Cruyff Foundation ad left.

ANNOTATIONS

Cruvff Foundation Logos

Labels human action person produc rano presentation

CRUYFF BRENGT JEUGO IN BEWEGING Nu al ruim 50 watertappunten op Cruyff Courts dankzij de deelnemers van de Nationale Postcode Loterij. POSTCODE HLOTERIU Text CRUYFF BRENGT JEUGO BEWEGING Nu 50 al ruim watertappunten Cruvff Courts deelnemers Nationale Postcode Loterii POSTCODE HLOTERIU

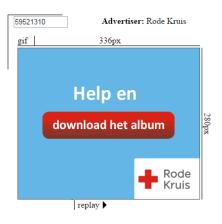
ue: 182 green: 65, red. 9 blue: 16 green: 229, red: 241 blue: 35 green: 94, red: 73 blue: 52 green: 195, red: 226 blue: 26 green: 57, red: 40 blue: 20 green: 148, red: 99 blue: 249 green: 200, red: 102 Colors e: 144 green: 85, red: 25 blue: 156 green: 75, red: 3 blue: 253 green: 254, red: 254

Faces								
angerlikelihood	sorrowlikelihood	surpriselikelihood	joylikelihood	headwearlikelihood	confidence	pan angle	roll angle	tilt angle
VERY_UNLIKELY	VERY_UNLIKELY	VERY_UNLIKELY	VERY_LIKELY	VERY_UNLIKELY	0.999953	6.05152	1.96839	3.0981
VERY_UNLIKELY	VERY_UNLIKELY	VERY_UNLIKELY	LIKELY	VERY_UNLIKELY	0.999564	9.01966	0.992587	-10.8816

Print screen from the Mediabrands Marketing Sciences Ad Vision dashboard

CONCLUSIONS

- □ Size does matter: the bigger ad format the better the results
- □ Horizontal is the best shape
- $\hfill\square$ The best colors to use are blue and black
- □ A logo adds value and increases CTR
- □ The suggested position of the logo is bottom-left
- □ Either a few or a lot of **text** increases the chance on a high CTR
- □ A call to action works: 'win' or 'bekijk' work best
- □ Multiple CTA's are contra productive: it scores lower then one CTA





FRAME DATA

19026 1 NColors: 11

WWW.MEDIABRANDSADVISION.NL

Mediabrands Advision website

We created a website where you can test the probable effectiveness of your ad based on the learnings of our research.

How does it work?

First you select the characteristics of your ad: product category, size, color, logo position (if any), call to action type (if any) and type of content that is depicted in the site. Select this by

At the bottom of the page you will see the predicted CTR based on our data.

Full report

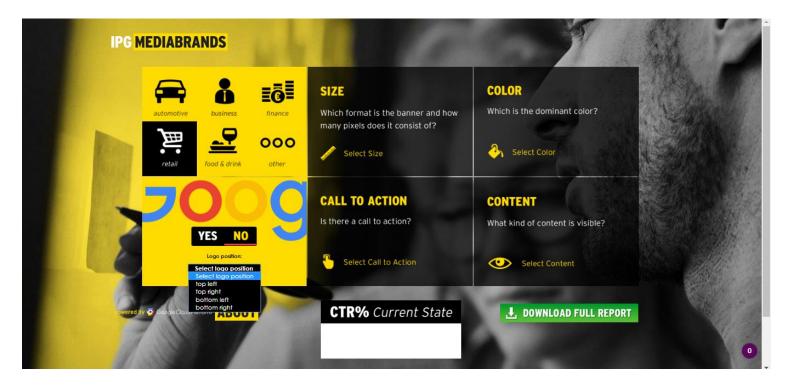
At the site you can download the full report of this project and read more about the medthodology.





http://www.mediabrandsadvision.nl

WWW.MEDIABRANDSADVISION.NL





THE IMPACT OF PRODUCT CATEGORY

The product category has a significant impact on the average CTR of an online ad. For benchmarking purposes we added the average for each category in this report. We indexed the CTR scores (100 is the average over all product categories). More information about average CTR's and other metrics can be found here: http://www.richmediagallery.com/tools/benchmarks.

In general, ads performed best in the following categories:

Eye care (index 269)

Energy and water suppliers (250)

Government and non-profit (225)

Food (general) (210)

Retail department stores (190)

	Category	subcategory	CTR index
1	Automotive	General	99
2	Business and industrial Business and industrial Business and industrial Business and industrial Clothing, shoes, accessories Drink and beverage	Business-to-business	48
3	Business and industrial	Employment, recruitment, staff	120
4	Business and industrial	General	39
5	Business and industrial	Information technology, high-tech	182
6	Clothing, shoes, accessories	General	72
10	Drink and beverage	Non-alcoholic, soft drinks	49
11	Education		37
	Financial services	Banks, credit cards, loans	54
	Financial services	General	34
	Financial services	Insurance	51
	Food	Fruit and vegetables	100
	Food	General	210
	Food	Snacks & Candy	93
	Government and non-profit	General	225
	Government and non-profit	Charities and voluntary organizations	162
	Household and domestic	General	138
	Leisure and entertainment	Hi-Fi, A/V, music players	22
	Leisure and entertainment	Lotteries, casinos, gambling	42
	Leisure and entertainment	Museums, attractions	69
	Media and publishing	General	175
	Pharmaceutical and healthcare		53
- ·	Retail	Clothing and fashion	114
	Retail	Department stores	190
	Retail	Electrical retailers	50
	Retail	Furniture and furnishings	47
	Retail	Online shopping, ecommerce	47
	Retail	Pharmacies and drug store	90
	Retail	Supermarkets and grocery stores	69
	Retail	Toys	22
	Retail	Eye care and ear care	269
46	Retail	General	38
49	Travel, transport and tourism		104
50	Utilities and services	Energy and water suppliers	250
	Utilities and services	General	140
53	None		96

EXAMPLES OF CORRECT LOGO IDENTIFICATION



OM VOORUIT TE KOMEN IN DE MAATSCHAPPIJ...

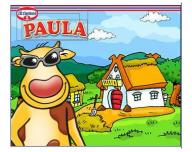




CRUYFF FOUNDATION / POSTCODE LOTERIJ



EXPERT / ZANUS



DR OETKI

We were amazed by the accuracy of the logo detection functionality of Vision API.

We were also surprised that Vision API didn't recognize the logo's on the next page.



BART SMIT / DESTIN

EXAMPLES OF WRONGLY IDENTIFIED LOGO'S





Rockwell **Automation**





LABEL ASSOCIATIONS & RECOGNIZING OBJECTS

The limits of Vision API

At the right side we see an ad to promote Koopmans 'pepernoten'. Vision API says 'food' (correct), 'product' (correct), plant, (strange association), mushroom (understandable since they look like it...

The Miele Vacuum cleaner is recognized as a steering wheel, which makes sense given the round shape of the tube. It is a home appliance but the other associations are wrong.



Produce	0,963
Food	0,933
Plant	0,899
Mushroom	0,795
Protea family	0,710

42
32
72
27
00



https://cloud.google.com/vision/

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LABEL ASSOCIATIONS

FRAME DATA



15289	0 NColors:	246

ANNOTATIONS								
Logos								
Labels folk dance								
Text								
	: 54, red: 42 blue: 39 green: 37 blue: 29 green: 41, red: 80		red: 57 blue: 125 green:	147, red: 197 blue: 10 green: 19,	red: 50 blue: 187 green	n: 212, red: 253 blu	ne: 55 green: 70, re	d: 117
Faces								
angerlikelihood	sorrowlikelihood	surpriselikelihood	joylikelihood	headwearlikelihood	confidence	pan angle	roll angle	tilt angle

RECOGNITION OF ANIMALS

FRAME DATA



32893 0 NColors: 236



55 B (12)		
32924	2 NColors:	234

Logos	Royal Canin
Labels	cat pet mammal animal british shorthair
Text	ROYAL CANIN Kijk ROYAL CANIN Kijk
	blue: 125 green: 144, red: 133 blue: 99 green: 93, red: 176 blue: 36 green: 52, red: 58 blue: 44 green: 52, red: 57 blue: 67 green: 85, red: 94 blue: 100 green: 118, red: 126 blue: 78 green. 87, red: 92 green: 123, red: 128 blue: 145 green: 153, red: 157 blue: 182 green: 188, red: 191

FACES

angerlikelihood	sorrowlikelihood	surpriselikelihood	joylikelihood	headwearlikelihood	confidence	pan angle	roll angle	tilt angle

ANNOTATIONS								
Logos Royal Car	iin							
Labels pug	pet mammal a	nimal dog						
Text ROYAL CA geven?	NIN Waarom zouden we z	e dan wel dezelfde voeding geven?	ROYAL	CANIN Waarom	zouden we	ze dan we	dezelfde	voeding
		f green: 99, red: 179 blue: 225 gree 10, red: 202 blue: 227 green: 228, r		: 115 green: 120, red: 122 b	olue: 47 green: 50, red: 51	blue: 82 green: 85, red.	: 86 blue: 20 green:	22, red: 23
Faces								
angerlikelihood	sorrowlikelihood	surpriselikelihood	iovlikelihood	headwearlikeliho	od confidenc	e pan angle	roll angle	tilt angle

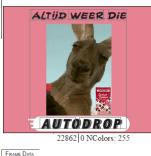
RECOGNITION OF ANIMALS



ANNOTATIONS	S											
Logos	South Africa											
Labels	mammal	animal	meerkat	vertebrate	carnivoran							1
Text	athAfri rica Ins	pringnewway	rs athAfri	rica	Inspringnewway	7S						
Colors	blue: 38 green:	152, red: 215	blue: 197 green	n: 232, red: 243	blue: 184 green:	166, red: 161 blue	e: 156 green: 193, red: 21	6 blue: 218 green	193, red: 179 bl	ue: 166 green: 156,	red: 155 blue: 62 g	green: 84, red: 100
blue: 106	green: 155, red:	189 blue: 87	green: 111, red	: 126 <mark>blue: 140</mark>	green: 192, red: 2	23						
	_											
Faces												
angerlik	elihood	sorrowlike	elihood	surpriseli	kelihood	joylikelihood	headwearlikel	ihood	confidence	pan angle	roll angle	tilt angle

RECOGNITION OF ANIMALS

FRAME DATA



ALTIJD WEER DIE



AUTODROP

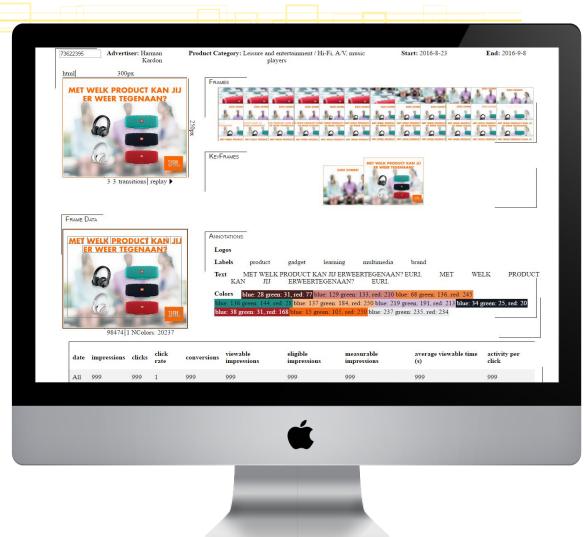
^{22920 0} NColors: 237

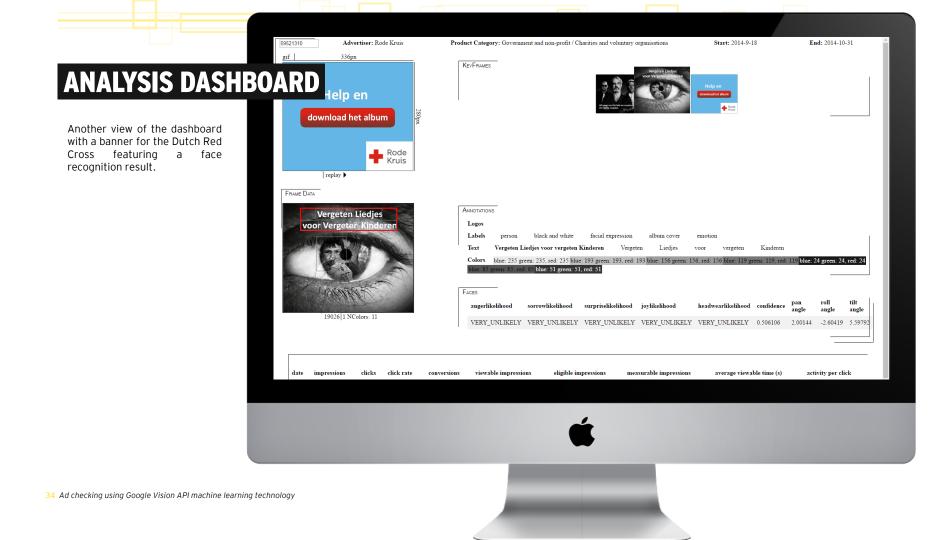
ANALYSIS DASHBOARD

To aid in the analysis we build a dashboard that displays the results of the project. In the image on the right a screenshot of the dashboard is shown displaying information on a single banner.

On the top general information about the banner like the advertiser, product category and size can been seen next to thumbnails of all the frames of the animation and the two keyframes that where analyzed by cloud vision.

Below that information about the second keyframe is shown, with the actual image on the left and Vision results on the right. On the bottom is a table with performance metrics specified per day. (performance metrics were altered for this image)

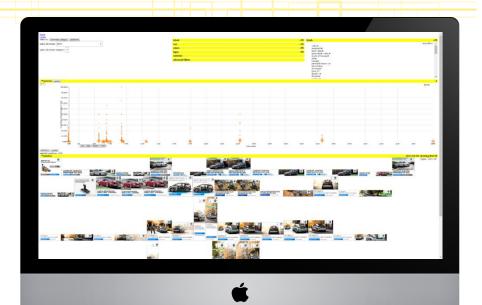




ANALYSIS DASHBOARD

Screenshots on the right show the dashboard displaying an aggregate view of selected banners.

At the top banners can be selected from the database based on the advertiser or other properties. Selected banners are displayed on the bottom. In the middle graphs can be shown with the CTR of the selected banners as a function of one of several properties (performance metrics were altered for this image).











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