

### More Low-Hanging Fruit Than You Think

Enabling non-Experts with Google ML Services

Kevin Kissell, Office of the CTO

Google Cloud

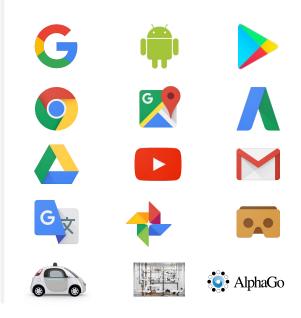


# Google is an Al first company





Used across products:





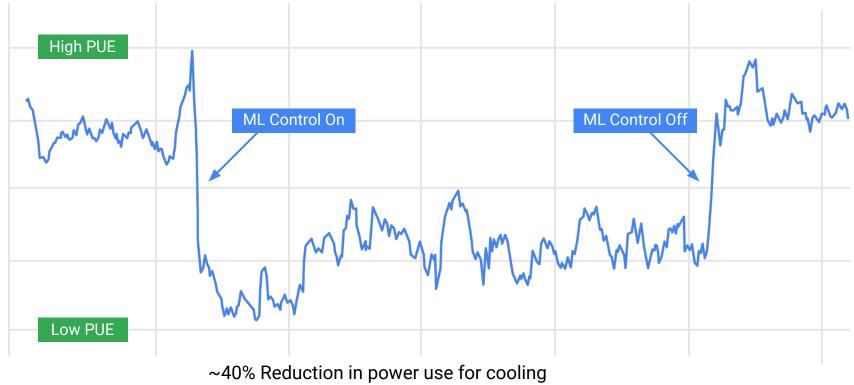
One of the largest server manufacturers. Zero servers sold.

Coogle Cloud



# "Eating Our Own Dog Food"

### Machine Learning to drive down PUE





### TPUs

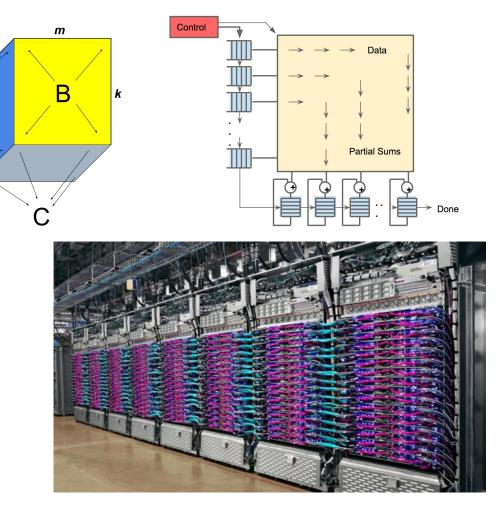
Matrix Multiplies Dominate ML Computation

Systolic Array Multiplier Architecture Maximizes Parallelism with Minimal Architectural State

> 100 TOPs per chip> 100 POPs per pod in a toroidal mesh

Support processors handle communication, perform JIT compilation on XLA from RPCs

Only programmable in TensorFlow and PyTorch





# **ML-enabled Astronomy**

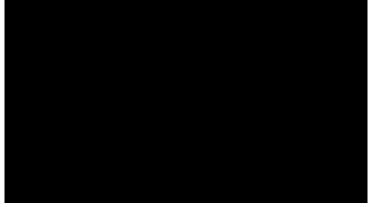
NASA/Kepler Mission Data

UT Austin and Google, trained deep learning network with known exoplanet transit data.

Discovered 2 New Exoplanets, Kepler 90g and Kepler 90i









### ML Predicting Weather

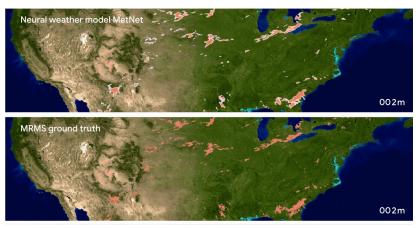
Google Brain Research for US NOAA

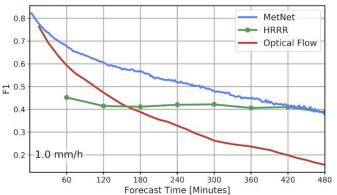
MetNet Neural Weather Model

Runs on 256 Google TPUs

Outperforms current physics based models for speed and accuracy out to 8 days

Parallel scaling allows prediction for entire US in seconds.





https://arxiv.org/abs/2003.12140



### AlphaFold: ML for a Scientific Grand Challenge

Prediction of 3D Protein Structure, based on "1D" molecular sequence data

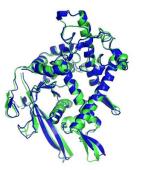
**DeepMind** used deep neural networks to predict Structure after training on ~170,000 known proteins

RMS Error of 0.1nm, scale of single atom

Comparable to results from experimental methods, but can be used for proteins not easily analysed.

This computational work represents a stunning advance on the protein-folding problem, a 50-year-old grand challenge in biology. It has occurred decades before many people in the field would have predicted. It will be exciting to see the many ways in which it will fundamentally change biological research.

> PROFESSOR VENKI RAMAKRISHNAN NOBEL LAUREATE AND PRESIDENT OF THE ROYAL SOCIETY



T1037 / 6vr4 90.7 GDT (RNA polymerase domain) **T1049 / 6y4f** 93.3 GDT (adhesin tip)

Experimental	result
Computational	prediction



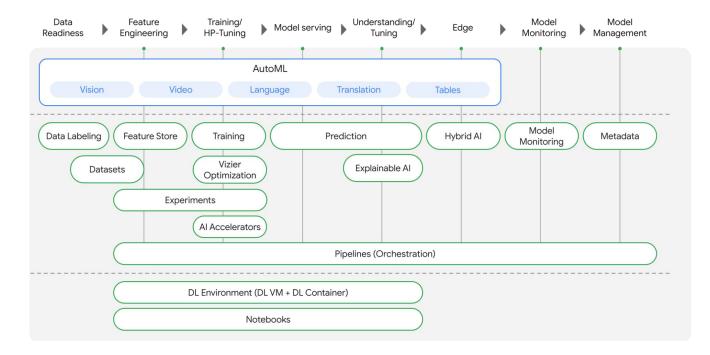
# ML must become easier and more accessible

Data Scientists vs. Developers (2018)									
Software Developers 📕 Data Scientists									
2300000									
	10000	00							
0	5000	0000	1000	0000	1500	0000	2000	00000	

**Google** Cloud

https://www.kdnuggets.com/2018/09/how-many-data-scientists-are-there.html https://www.quora.com/How-many-software-developers-are-there-in-the-world

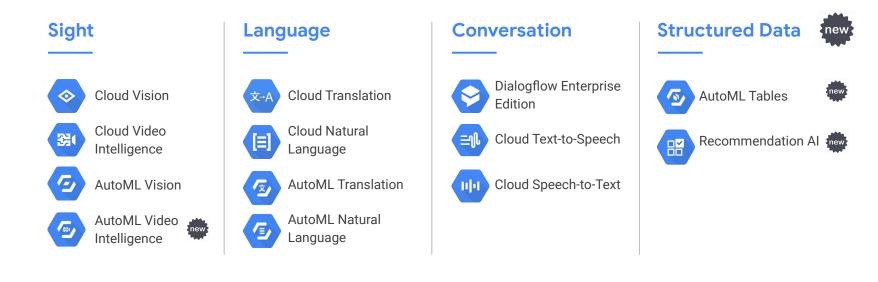






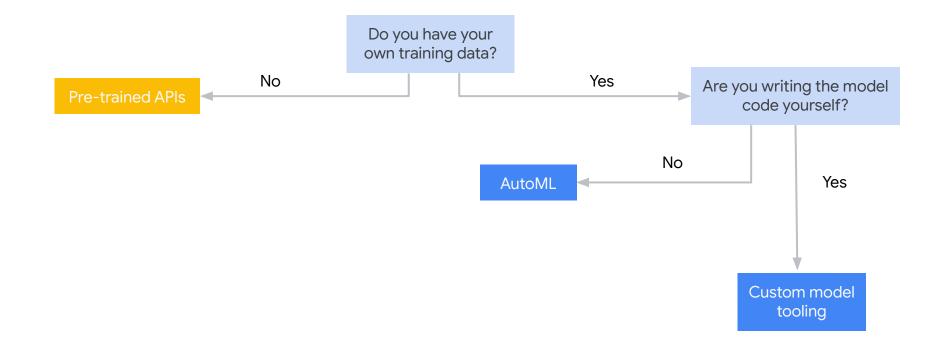
# **APIs and AutoML**

### Making ML accessible to all developers



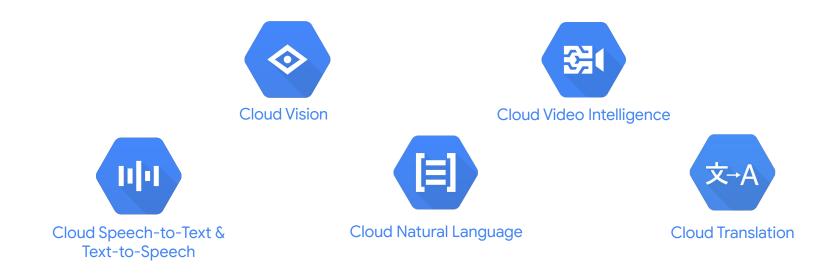


### What's the right tool for you?



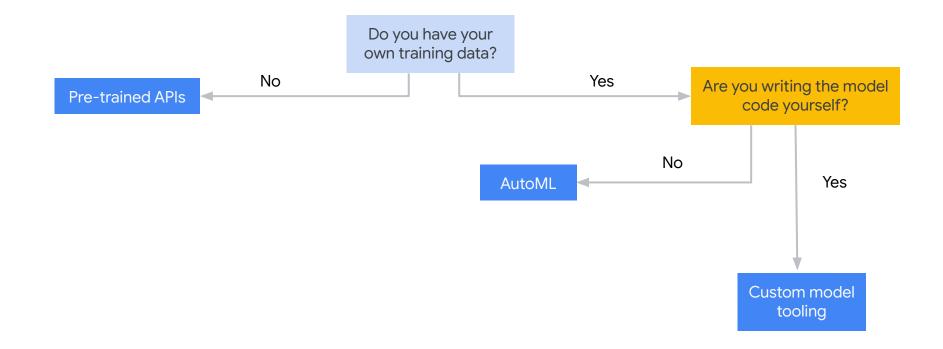


# Use a **pre-trained model** to accomplish common ML tasks





### What's the right tool for you?

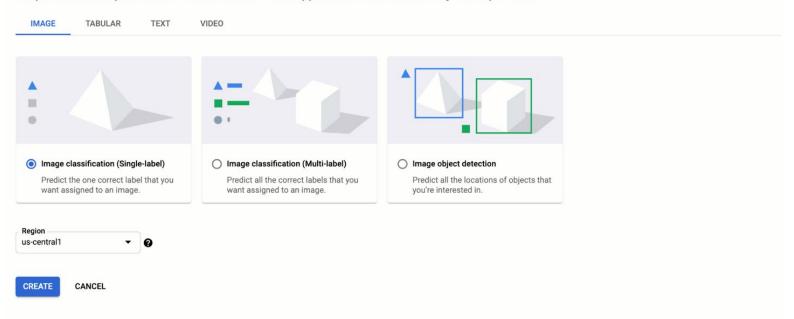




## Bringing your OWN Data? First choose your data type..

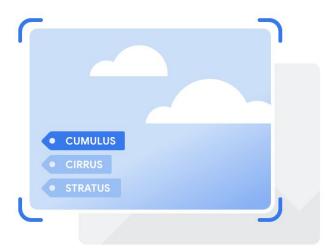
#### Select an objective

An objective is an outcome you want to achieve with a trained model. Don't worry, you can use this dataset for other image-based objectives later.

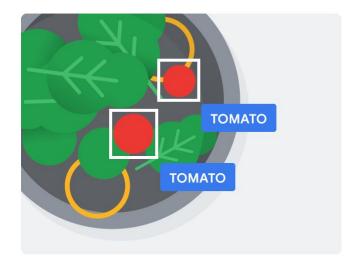




### Image datasets



**Image classification** models predict one (or many) labels for an image. For example, identifying types of clouds from images of the sky.



**Object detection** models draw bounding boxes around items of interest. For example, identifying vegetables from images of food.

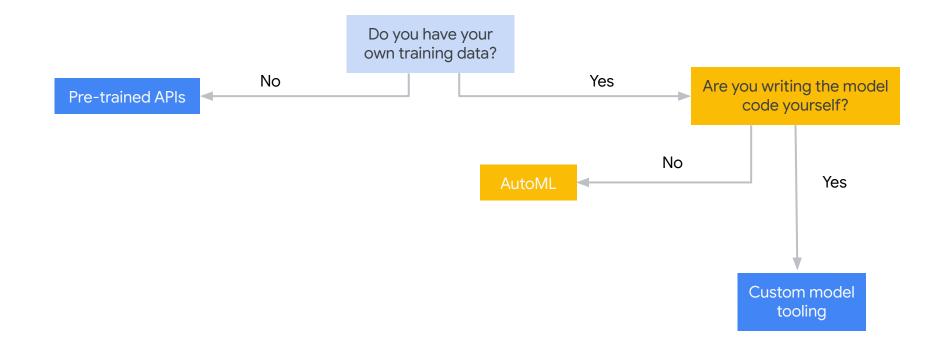


### Should you write your own model code? It depends.

Use AutoML if	Build a custom model if
Your use case fits into our supported AutoML offerings:	Your use case doesn't fit into any AutoML use cases
<ul> <li>Image &amp; video classification or object detection</li> <li>Text classification, entity extraction, sentiment</li> <li>Tabular regression, classification, or forecasting</li> </ul>	OR Your model takes mixed input types, like images + tabular metadata
You don't need to know specifics about the underlying model	You want control over your model's architecture, framework , or exported model assets. For example, maybe your model needs to be built with TensorFlow
You want to develop a quick initial model to use as a baseline (which could end up being your production model)	You already have a baseline or heuristic and you want to see if you can improve upon it

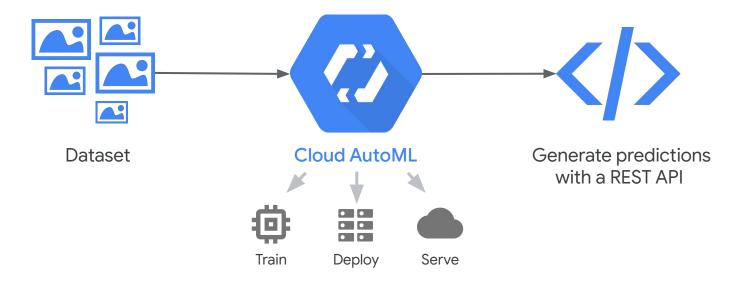


### What's the right tool for you?





# AutoML ML that creates ML



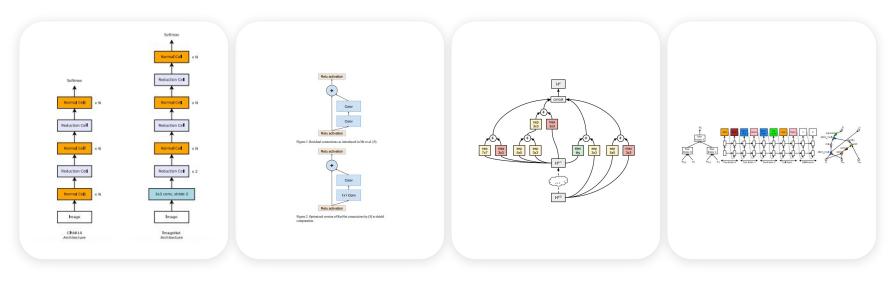


### **Cloud AutoML - Best in Class Research**

#### Learning to learn

**Transfer Learning** 

#### Hyperparameter Tuning



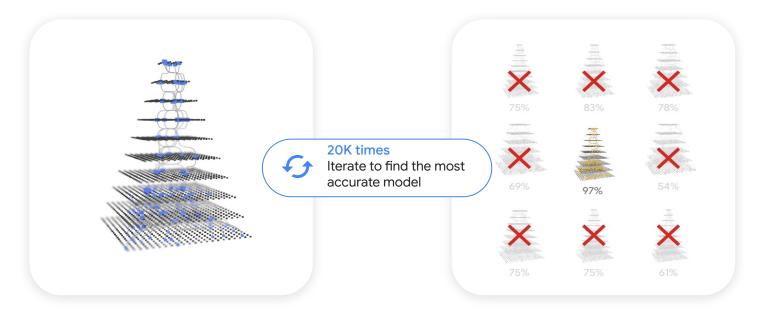
Learning Transferable Architectures for Scalable Image Recognition, Barret Zoph, Vijay Vasudevan, Jonathon Shlens, and Quoc V. Le. Arxiv, 2017. Inception-v4, Inception-ResNet and the Impact of Residual Connections on Learning Christian Szegedy, Sergey loffe, Vincent Vanhoucke, and Alex Alemi. AAAI, 2017. Progressive Neural Architecture Search Chenxi Liu, Barret Zoph, Jonathon Shlens, Wei Hua, Li-Jia Li, Li Fei-Fei, Alan Yuille, Jonathan Huang, Kevin Murphy, Arxiv, 2017 Neural Architecture Search with Reinforcement Learning Barret Zoph, Quoc V. Le. ICLR 2017.



### How does AutoML work?

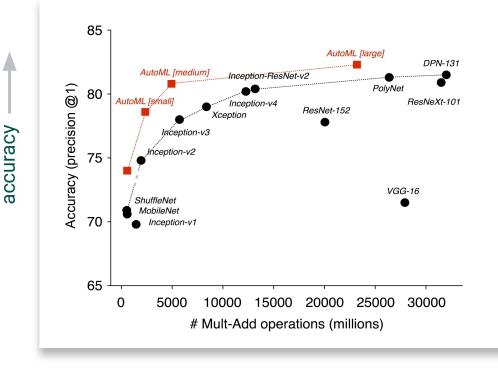
#### Controller: proposes ML models

#### Train & evaluate models





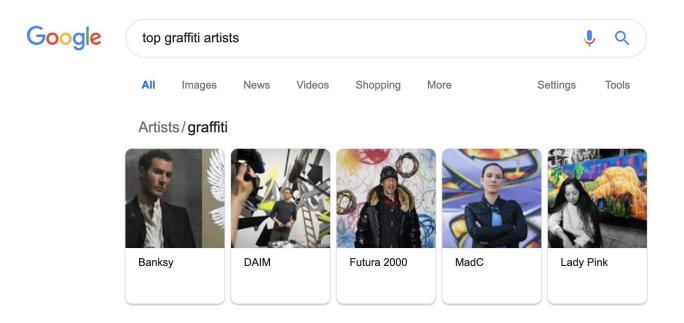
## AutoML outperforms handcrafted models



computational cost

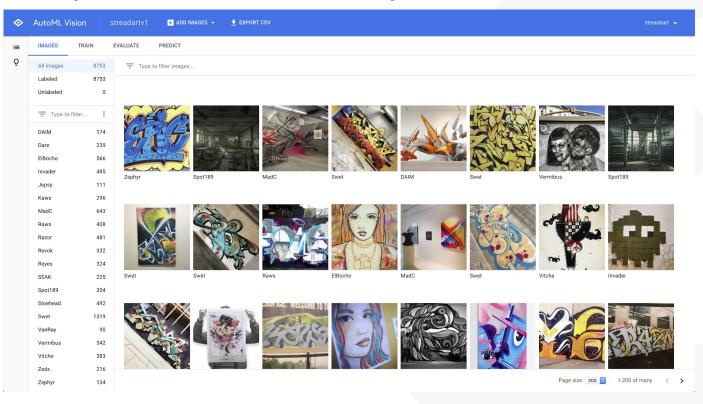


Learning Transferable Architectures for Scalable Image Recognition, Zoph et al. 2017, https://arxiv.org/abs/1707.07012





# AutoML Step 1: Show me examples





# AutoML Step 2: Click a button and wait

	AutoML Vision	streadartv1	+ ADD IMAGES		RT CSV					
:=	IMAGES TRAIN	EVALUATE	PREDICT							
Q	Models TRAIN NEW MODEL TRAIN ADVANCED MODEL streadartv1_201806010506_advanced									
	CreatedAnalyzedAvg precision ③ 0.956Precision 94.0%Jun 02, 2018 12:37 PM8753 images 26 labels, 935 test images0.95694.0%									
	Precision and recall are based on a score threshold of 0.5 SEE FULL EVALUATION									



# AutoML Step 3: Evaluate

AutoML Vision stre	adartv1 🕂 ADD IMAGES 👻 👱 EXPORT CSV			streadsart 👻
IMAGES TRAIN EVA	UATE PREDICT			
Model streadartv1_201806010506_advanced	B			
Created Jun 02, 2018 12:37 PM	Analyzed Avg precision 8753 images 0.956 26 labels, 935 test images	⑦ Precision ⑦ 94.0%	Recall ⑦ 90.0%	
Precision and recall are based on a sc	re threshold of 0.5			
Type to filter labels				
All labels	All labels			
Atome	Score threshold ⑦ 0.50			
Banksy	Total images 8753			
Bates	Precision ⑦ 94.1%			
BlekLeRat	Recall ② 90.2%			
CanTwo				
CaseWasHere	1.00	1.00	1.00	
DAIM				
Dare	0.75	0.75	0.75	
ElBocho	5 3 0.50	ඟ දු 0.50	0.50	
Invader	Pre	a,	and the second se	
Jepsy	0.25	0.25	0.25	
Kaws	0.00	0.00	0.00	
MadC	0.0 0.2 0.4 0.6 0.8 1.0	0.0 0.2 0.4 0.6 0.8 1.0	0.0 0.2 0.4 0.6 0.8 1.0	
Raws	Recall	Score threshold	Score threshold	
Razor				

C Google Cloud

# AutoML Step 3: Evaluate

#### AutoML Vision > Datasets > streadartv1 streadsart Confusion matrix ~ Predicted label BlekLeRat Banksy VanRay Zephyr nvader Reyes MadC Vitche Swet Dare Banksy 0% 5% 0% 0% 2% 2% 2% 2% 0% BlekLeRat 11% 0% 0% 0% 88% 0% 0% 0% Vitche 0% 0% 0% 0% 100% 0% 0% 0% 0% 0% Swet 1% 0% 0% 0% 98% 0% 0% 0% 0% 0% Zephyr 0% 0% 85% 0% 0% 0% 7% 0% 0% 7% VanRay 0% 0% 100% 0% 0% 0% 0% 0% 0% 0% Reyes 0% 0% 8% 0% 0% 0% 0% 0% 86% 4% 0% 0% MadC 0% 0% 0% 0% 0% 94% 0% 5% **Frue Label** Invader 0% 2% 0% 0% 0% 0% 0% 0% 97% 0% 3% 0% 0% 3% 3% 3% 0% 10% 0% Dare 75%

**Google** Cloud

# AutoML Step 3: Evaluate



	e	
COMMENT	A	

	Banksy	BlekLeRat	Vitche	Swet	Zephyr	VanRay	Reyes	MadC	Invader	Dare
Banksy	82%	0%	5%	0%	0%	2%	2%	2%	2%	0%
BlekLeRat	11%	88%	%	0%	0%	0%	0%	0%	0%	0%
Vitche	0%	0%	100%	0%	0%	0%	0%	0%	0%	0%
Swet	1%	0%	0%	98%	0%	0%	0%	0%	0%	0%
Zephyr	7%	0%	0%	0%	85%	0%	0%	0%	0%	7%
VanRay	0%	0%	0%	0%	0%	100%	0%	0%	0%	0%
Reyes	0%	0%	0%	0%	0%	0%	86%	8%	0%	4%
MadC	0%	0%	0%	0%	0%	0%	0%	94%	0%	5%
Invader	0%	2%	0%	0%	0%	0%	0%	0%	97%	0%
Dare	3%	0%	0%	3%	3%	3%	0%	10%	0%	75%

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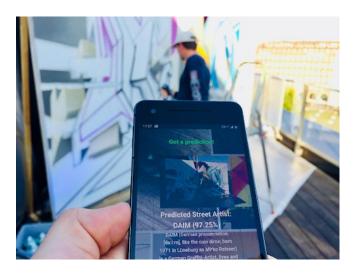
Predicted label

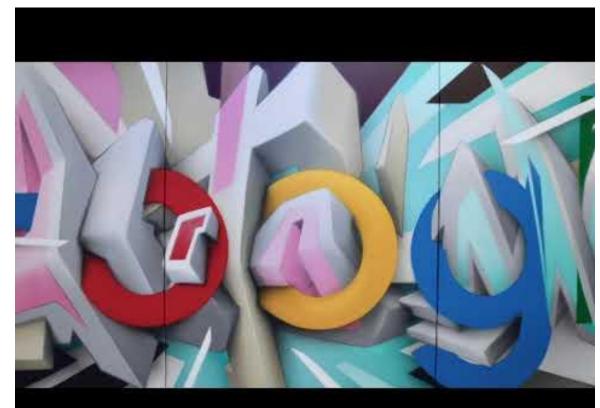


True Label



# AutoML Step 4: Use the Model (streetartist.app)







# **AutoML Tables** Go from raw data to models in days instead of months

Produce state-of-the art models with one click

Guides users through the full machine learning lifecycle without code

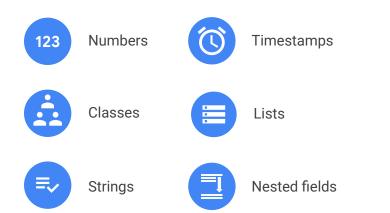
Uses Google's serving infrastructure making deployment fast and easy





### Handle data as found in the wild

### Automated feature engineering for:



### **Resilient to + guardrails for:**



Imbalanced data



Highly correlated features



Missing values



High cardinality features (like IDs)



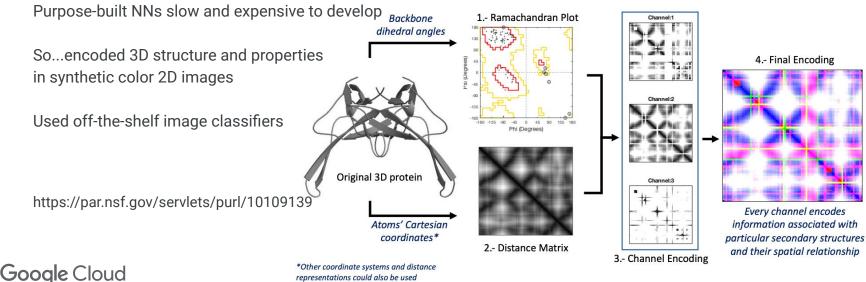
Outliers



### **Back to Proteins: Imagination Before Brute Force**

#### Reduce to a Solved Problem

Estrada et. al. at UNM wanted to identify proteins automatically



representations could also be used

# The Machine Learning Use Cases are there (so is the data)



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### Manufacturing

- Predictive maintenance
- Energy efficiency
- Warranty reserve estimation
- Propensity to buy
- Demand forecasting
- Process optimization
- Telematics



- Aircraft scheduling
- Dynamic pricing
- Social media consumer feedback and interaction analysis
- Customer complaint resolution
- Traffic patterns and congestion
   management



- Predictive inventory planning
- Recommendation engines
- Upsell and cross-channel marketing
- Market segmentation and targeting
- Customer ROI and lifetime value

### **Financial Services**

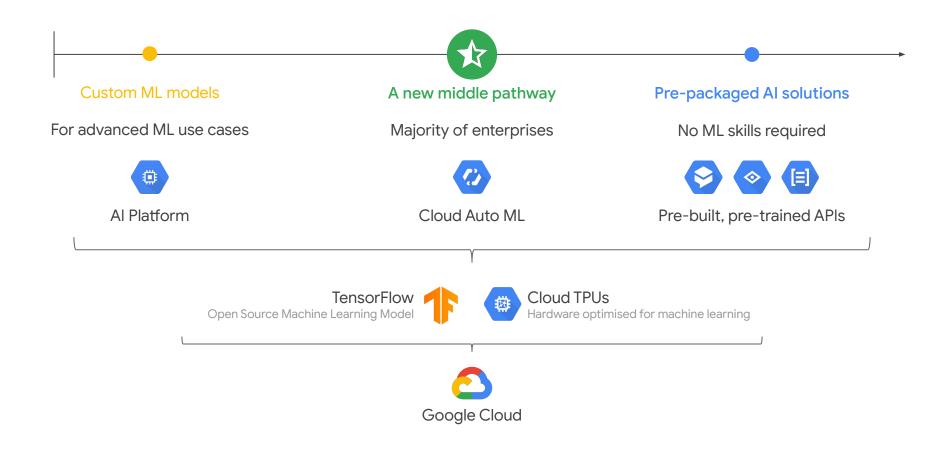
- Risk analytics and regulation
- Customer Segmentation
- Cross-selling and up-selling
- Sales and marketing campaign management
- Credit worthiness evaluation

#### Healthcare and Life Sciences

- Alerts and diagnostics from real-time patient data
- Disease identification and risk
   stratification
- Patient triage optimization
- Proactive health management
- Healthcare provider sentiment
   analysis

### Energy, Utilities and Raw Materials

- Power usage analytics
- Seismic data processing
- Carbon emissions and trading
- Customer-specific pricing
- Smart grid management
- Energy demand and supply optimization





# Thanks for Your Attention!

And think about the ways that Google Cloud could become part of YOUR lab equipment...



