

Kolmogorov Arnold Networks

Hou, Y., Zhang, D., (2024). **A Comprehensive Survey on Kolmogorov Arnold Networks (KAN)**

Overview

- ▶ Kolmogorov (RUS) & Arnold (UKR) have posited that any continuous multivariate function (*one in which a small variation in function implies a small variation in result*), can be represented by a finite number of univariate functions (*only one variable*), and those univariates can replace the weight parameters of a neural network.
- ▶ The theorem, proposed in 1957, is based on several mathematical theorems involving numerical analysis and partial differential equations.
- ▶ The advantages are improvements in data fitting and complex learning tasks, as well as performance with high dimensional data in their pursuit of capturing complex temporal dependencies.

Key Differences: Standard MLP v. KAN

MLP

- ▶ Goal is to optimize based on learned parameters w & b
- ▶ Predefined activation function

KAN

- ▶ Goal is to optimize based on learned coefficients of spline, with unfixed functions
- ▶ Flexible activation function (spline)

Main Goal of KAN: Kolmogorov-Arnold Networks (KAN) enhance their ability to handle nonlinear relationships by using spline functions to implement complex nonlinear transformations on each edge.

Mathematical Expression of Theorem

$$f(x_1, x_2, \dots, x_n) = \sum_{i=1}^{2n+1} g_i \left(\sum_{j=1}^n h_{ij}(x_j) \right)$$

Where:

| | | |
|-----|---|---------------------|
| x | = | sample data |
| n | = | dimensions |
| g | = | univariate function |
| h | = | univariate function |

$$S_i(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2 + d_i \cdot (x - x_i)^3$$

Where:

| | | |
|-------|---|---------------------------------|
| x | = | any value along spline interval |
| x_i | = | knot point |
| $a-d$ | = | coeffs |

[Reference](#)

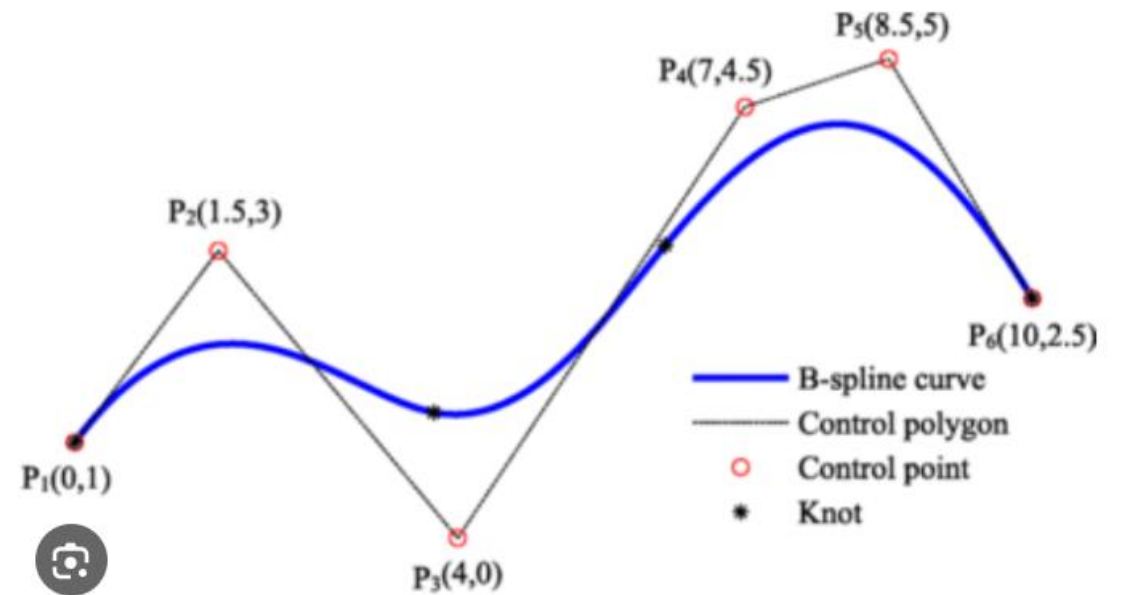
Mathematical Expression In Practice

$$B_{i,0}(x) = 1, \text{ if } t_i \leq x < t_{i+1}, \text{ otherwise } 0,$$
$$B_{i,k}(x) = \frac{x - t_i}{t_{i+k} - t_i} B_{i,k-1}(x) + \frac{t_{i+k+1} - x}{t_{i+k+1} - t_{i+1}} B_{i+1,k-1}(x)$$

where,

| | | |
|---|---|--------------------|
| x | = | data point |
| t | = | knots |
| k | = | exponential degree |

[Reference](#)



[Reference](#)

Basic MLP Example

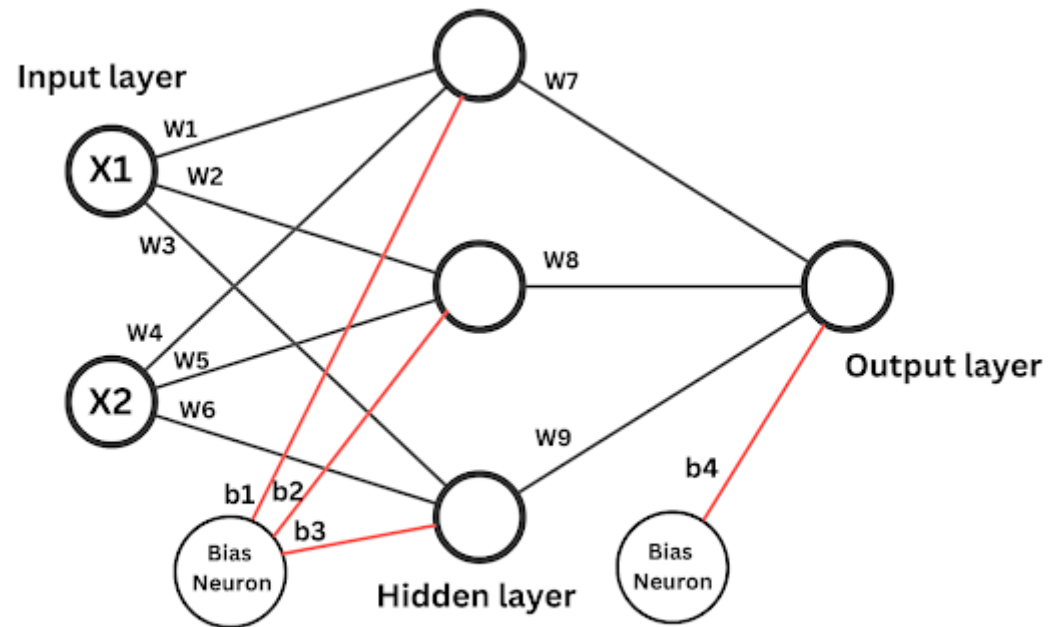


Fig 4 (MLP Weights and Biases)

[Reference](#)

Model Parameter Comparison

Basic Neural Network

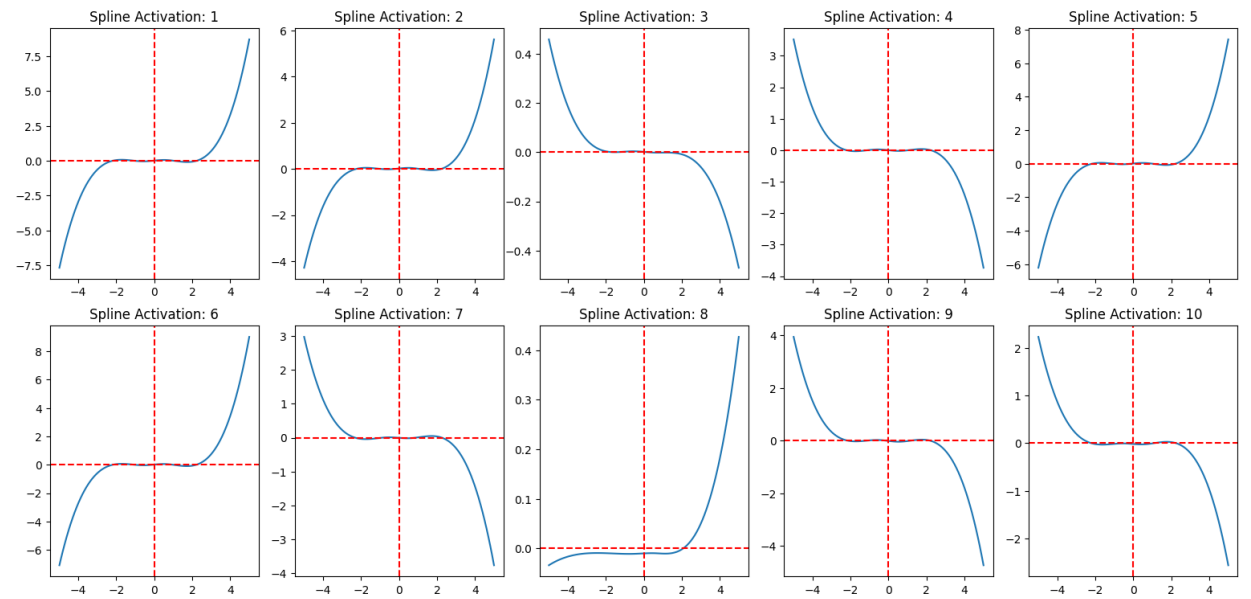
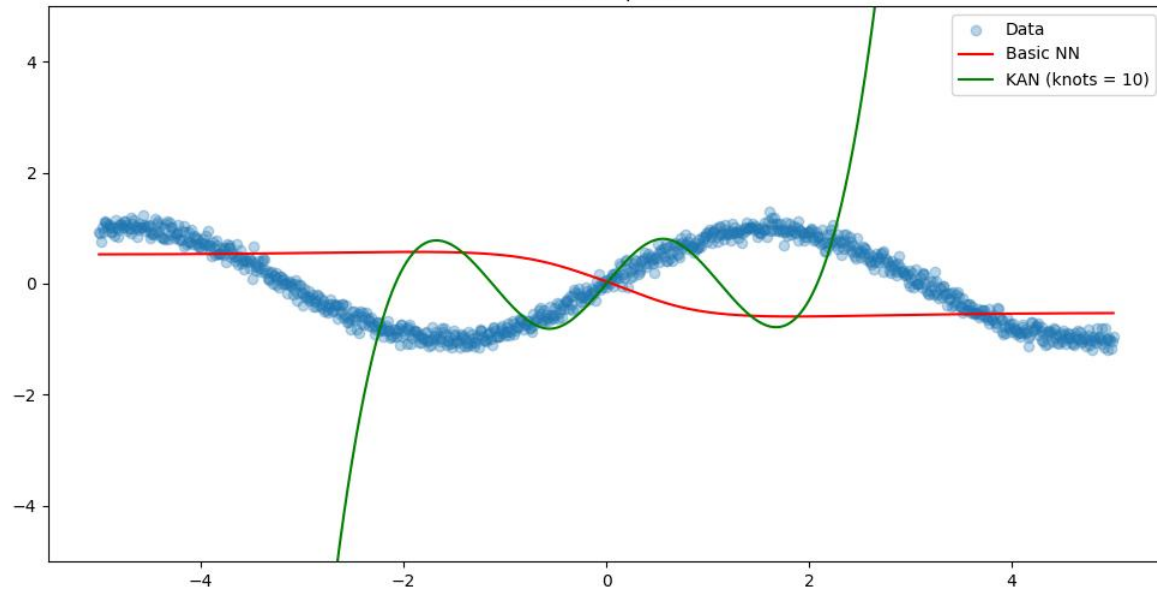
```
['w1', 'b1', 'w2', 'b2']  
[[ 0.60856746  0.94646912  1.36256079 -0.80811515  1.68428053  1.45438326  
  0.25701588  0.10033642  1.27488093  1.49932301]]  
[[-0.00685603  0.00436755 -0.00049083 -0.03636161  0.00264737 -0.00041073  
  0.00596033  0.09783393  0.00030786  0.00086073]]  
[[ 0.17837218]  
 [-0.18783807]  
 [ 0.26433984]  
 [ 0.38592123]  
 [ 0.4928461 ]  
 [-0.32040121]  
 [-0.4879605 ]  
 [ 0.69860776]  
 [-0.17076239]  
 [ 0.38719292]  
 [[-0.05554404]]
```

KAN

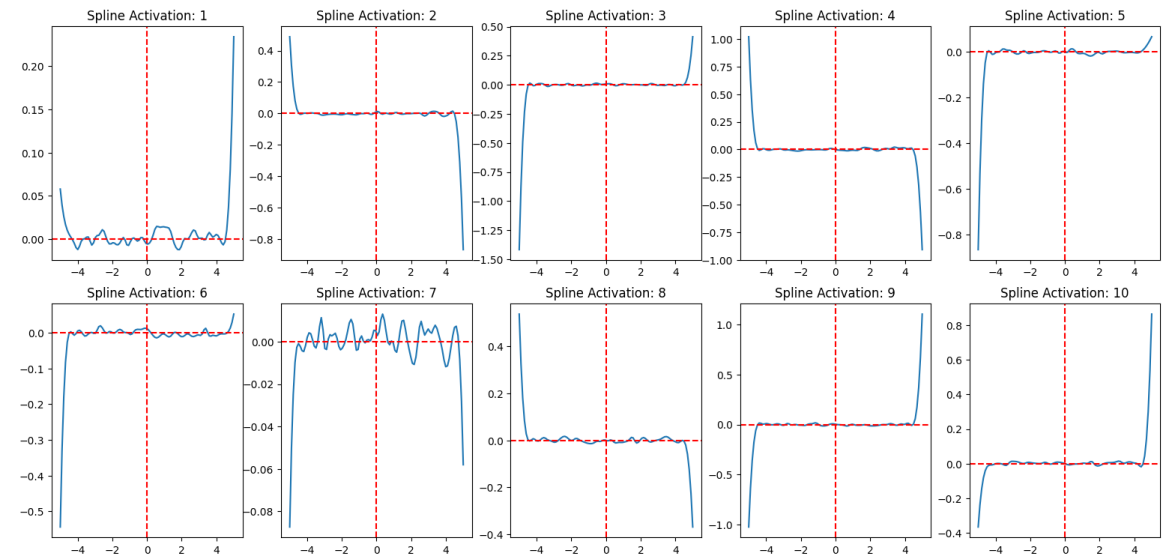
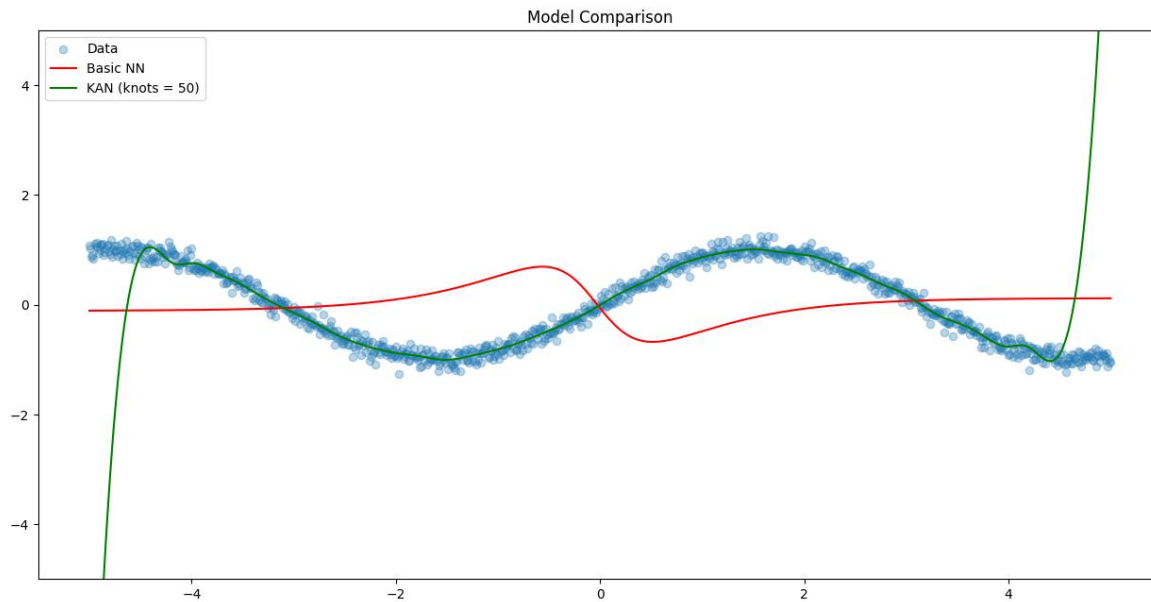
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['knots', 'coeffs', 'w', 'b']  
[-5. 1.66666667 2.77777778 3.88888889 5. 0.55555556 0.55555556  
 1.00000000e+00 -1.00000000e+00 1.00000000e+00 -1.00000000e+00  
 1.00000000e+00 -1.00000000e+00 8.21284089e-03 2.03309747e-03  
 -1.52311071e-02 -1.51004496e-04]  
[ 9.48938837e-01 -1.00000000e+00 1.00000000e+00 -1.00000000e+00  
 1.00000000e+00 -9.48908774e-01 1.14880059e-03 -8.31089032e-03  
 -3.74943916e-03 4.69016505e-03]  
[ 2.16978030e-01 -2.14890994e-01 -2.14332426e-01 2.14545746e-01  
 2.14893720e-01 -1.71872616e-01 4.53034001e-03 -1.14413120e-02  
 6.18545797e-03 7.43075008e-03]  
[-5.97341235e-02 -1.86845850e-01 -1.74881709e-01 1.63154871e-01  
 2.12948328e-01 9.64298628e-02 -8.49678504e-04 -6.23140917e-03  
 2.57250080e-02 -1.82942266e-03]  
[ 1.00000000e+00 -1.00000000e+00 1.00000000e+00 -1.00000000e+00  
 1.00000000e+00 -1.00000000e+00 2.22294817e-03 -9.96355659e-03  
 -1.87966794e-03 1.78938473e-02]  
[ 1.33624523e-02 1.42450332e-02 -9.94085883e-03 -2.07696589e-02  
 8.77440278e-03 6.82837422e-03 -3.74824117e-03 -1.70008327e-02  
 -7.39738265e-03 -1.82690510e-02]  
[-1.00000000e+00 1.00000000e+00 -1.00000000e+00 1.00000000e+00  
 -1.00000000e+00 1.00000000e+00 7.03913329e-03 -6.88053452e-03  
 -1.60209847e-02 4.53066129e-03]  
[-1.00000000e+00 1.00000000e+00 -1.00000000e+00 1.00000000e+00  
 -1.00000000e+00 1.00000000e+00 3.25165409e-03 -1.14205974e-03  
 -1.24350893e-02 1.43648394e-02]  
[-2.63353093e-03 -7.15177796e-03 -9.84193566e-03 -8.93985862e-03  
 2.21769810e-03 5.57305592e-04 -1.13821713e-02 2.63695837e-03  
 7.93932149e-03 2.91782974e-03]  
[-1.00000000e+00 1.00000000e+00 -1.00000000e+00 1.00000000e+00  
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 -4.01853884e-03 -1.71115533e-03]]  
[[-6.39517724e-01]  
 [-1.68538399e-01]  
 [ 3.62556624e-02]  
 [ 2.21791714e-02]  
 [-7.12217681e-01]  
 [-3.15505350e-06]  
 [ 5.80088044e-01]  
 [ 7.19485974e-01]  
 [ 5.49807206e-06]  
 [ 6.53063463e-01]]  
[[-0.00307258]]
```

Graphical Display @ 10 knots

Model Comparison

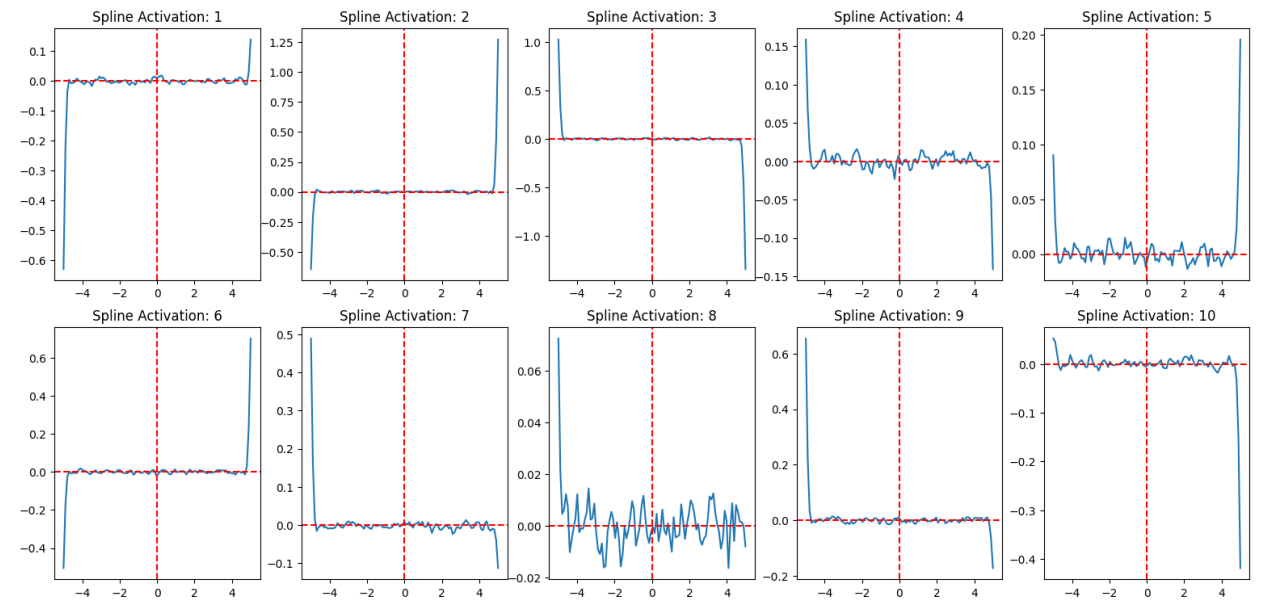
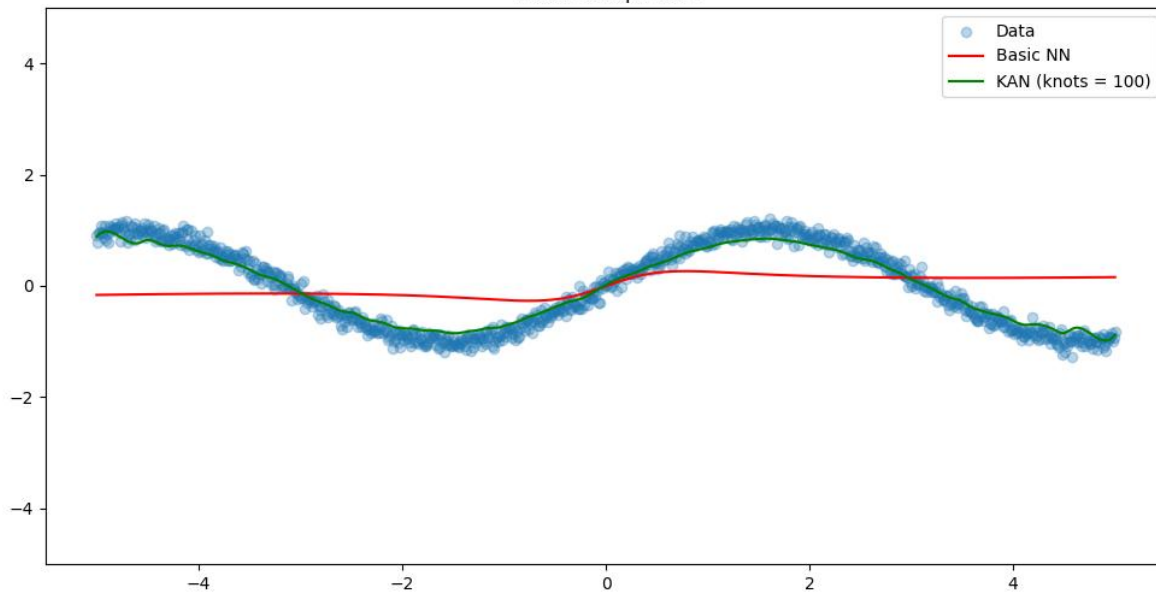


Graphical Display @ 50 knots



Graphical Display @ 100 knots

Model Comparison



Advantages & Uses for KAN

Advantages

- ▶ Fewer parameters than NN, in theory?
- ▶ Can adapt better to high dimensional data
- ▶ Better accuracy and interpretability

Uses

- ▶ Symbolic Regression (solving for formula)
- ▶ Time Series Prediction
- ▶ Graph-Structured Data Processing

Source Code

▶ [Google Drive](#)

▶ Questions?