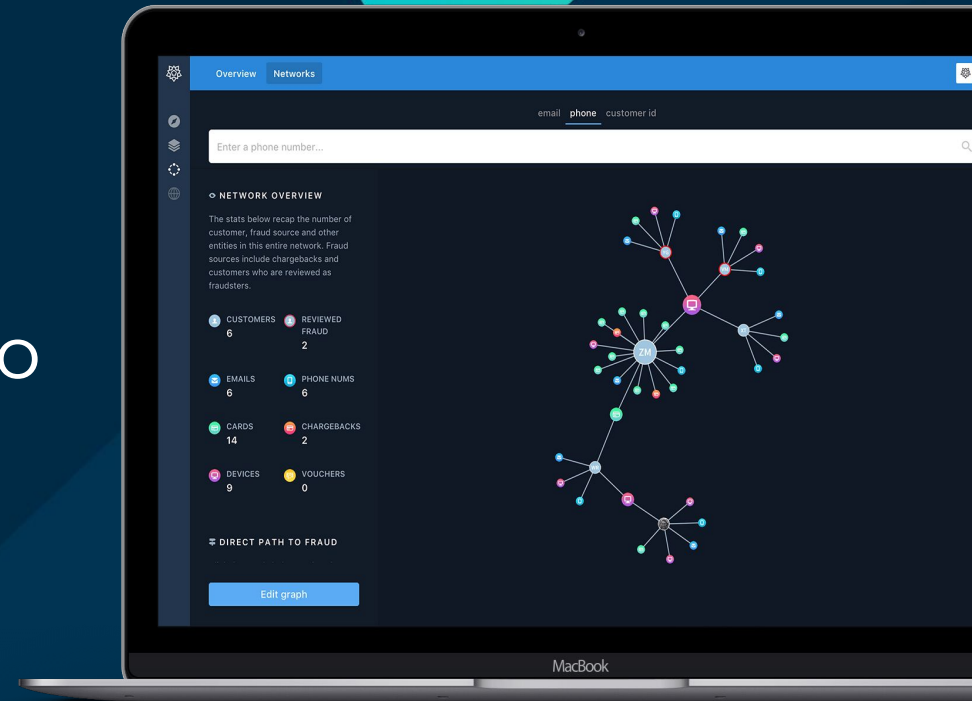


Applying software engineering principles to ML model deployments



Customers

Filter customers - Start typing to search for customers

15 customers of 110,259 1-15 of 110,259

SCORE	ID	NAME	EMAIL	ACCOUNT AGE	LAST ACTIVE	CARDS	ORDERS	SPENT	WATCHED
100	504661	Mame Murphy	shellie.cruickshank@gmail...	7mo	6mo	8	-	0.00	
100	538991	Madelynn Torp	mayert.ian@gmail.com	5mo	5mo	8	5	2,110.00	
100	358932	Amiyah Morar	harris.ireland@gmail.com	5mo	4mo	8	1	420.00	
100	357035	Mont Balistreri	lanie17@yahoo.com	8mo	6mo	8	1	419.00	
100	521598	Brandee	almeda34@hotmail.com	4mo	4mo	8	2	652.00	
100	487273	Dylon Bartoletti	emma.stracke@hotmail.com	9mo	7mo	8	-	0.00	
100	796054	Rollin	glenna.vonrueden@hotma...	4mo	4mo	8	1	528.00	
100	728997	Lavelle	hyatt.derrell@yahoo.com	6mo	4mo	9	2	494.00	
100	727199	Syed Heidenreich	mblick@hotmail.com	7mo	6mo	8	3	1,316.00	
100	218236	Monna Franecki	neha.price@gmail.com	4mo	4mo	8	2	720.00	
100	545766	Benji	feil.harding@yahoo.com	4mo	4mo	8	1	468.00	

About Us

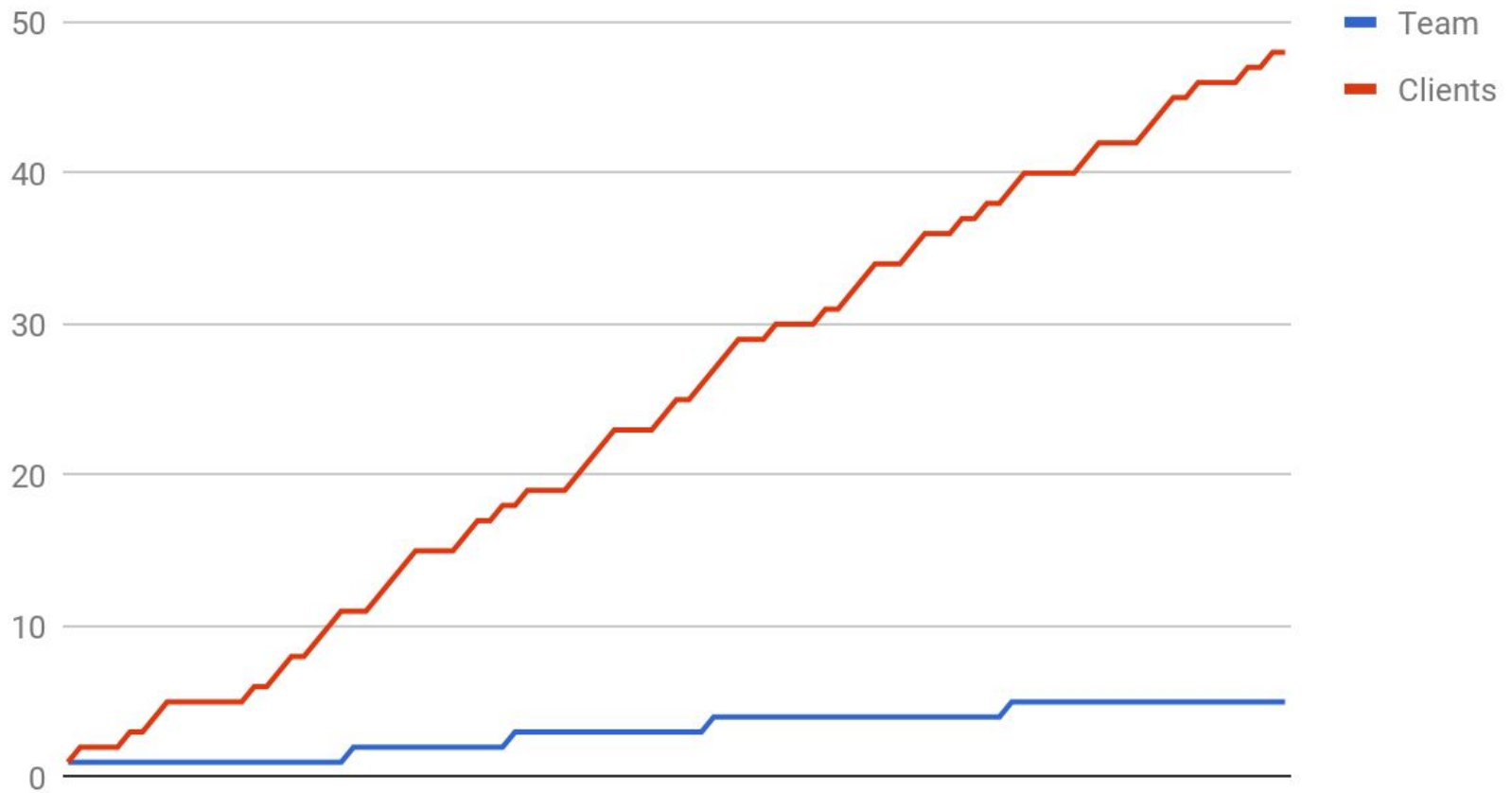
- B2B fraud detection
- Large client portfolio
- Each one special

Score customers in real time
for likelihood of fraud

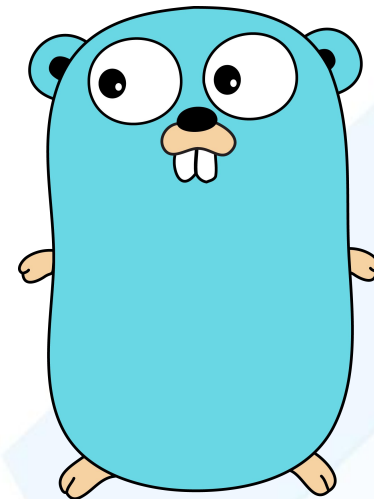
A bit of history

Move fast, don't break stuff

Team and client growth



How can we scale model
deployments without
scaling people?



A happy path is a default scenario featuring no exceptional or error conditions.

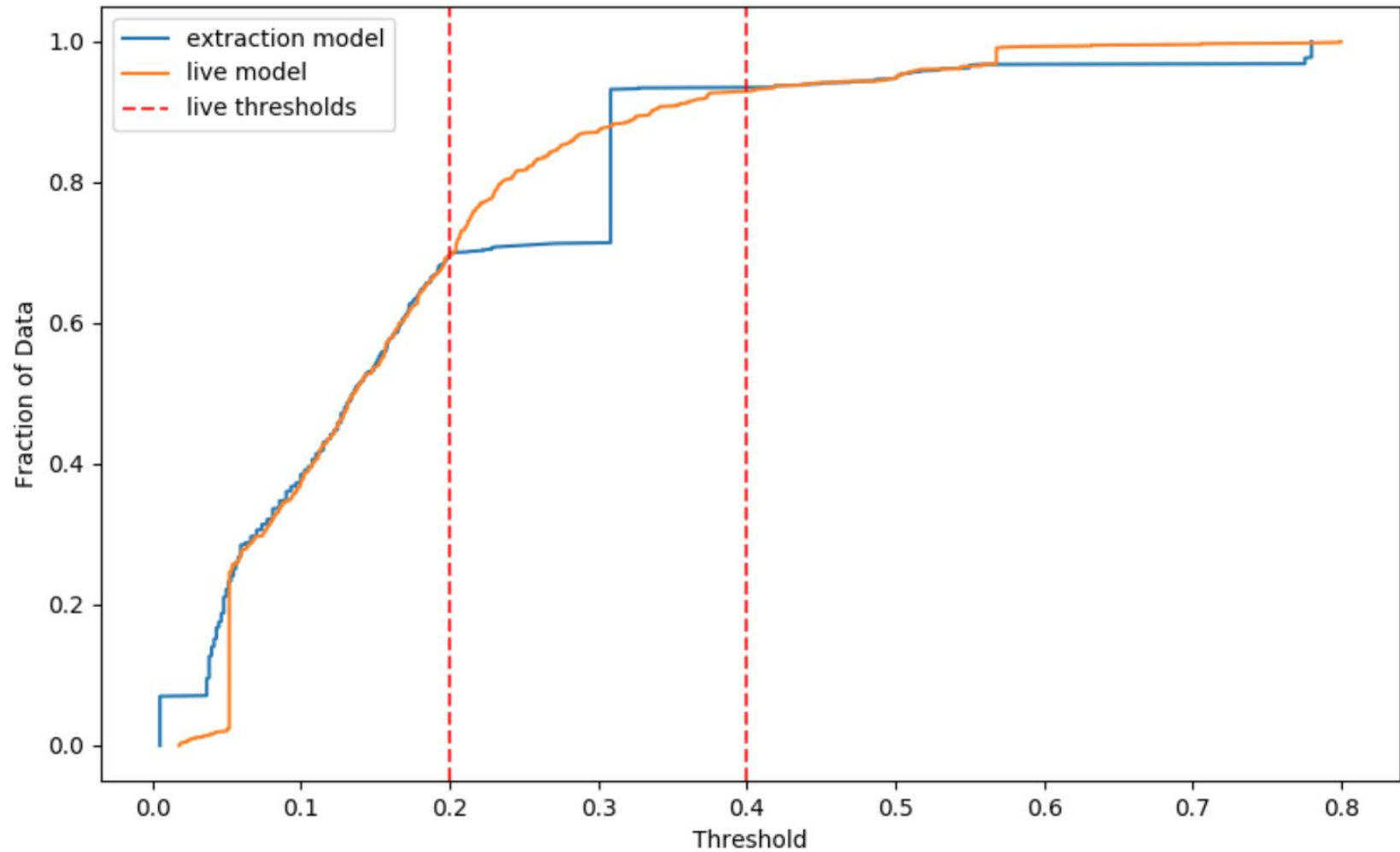
Happy Path (✗)

metric	extraction	live	status
test prauc	0.4664	0.4579	✓
prevented asserts	0.8000	0.7000	✓
calibration diff	0.0344	-	✓
feature rank	0.9493	-	✓
action changes	0.0707	-	✗

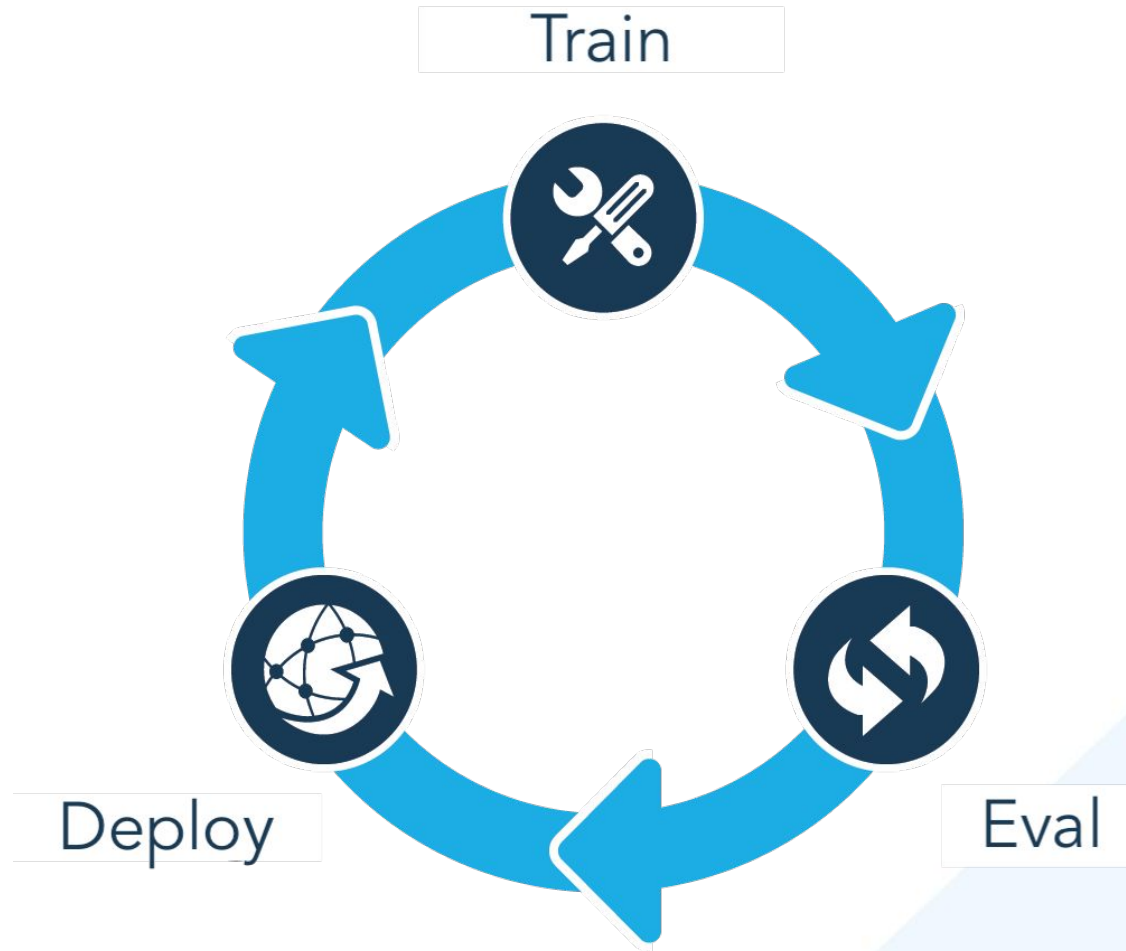
CSV

Search:

customerID	score	live score	abs score diff	action	live action	
	26	19	7	REVIEW	ALLOW	+
	53	73	20	PREVENT	PREVENT	+
	18	13	4	ALLOW	ALLOW	+
	45	55	10	PREVENT	PREVENT	+
	44	45	0	PREVENT	PREVENT	+
	21	32	10	REVIEW	PREVENT	+
	51	44	7	PREVENT	PREVENT	+
	86	76	9	PREVENT	PREVENT	+
	52	57	5	PREVENT	PREVENT	+
	13	14	0	ALLOW	ALLOW	+
	45	24	20	PREVENT	REVIEW	+
	46	39	6	PREVENT	PREVENT	+



Continuous integration



*Frequency reduces difficulty,
if something is hard, do it
more often. (Martin Fowler)*

once per quarter



once per week

Underfit using simple features,
simple models and strong
regularisation.

How can we scale onboarding new clients when each require a bespoke model?

Micromodels

Heterogeneous

Scalable

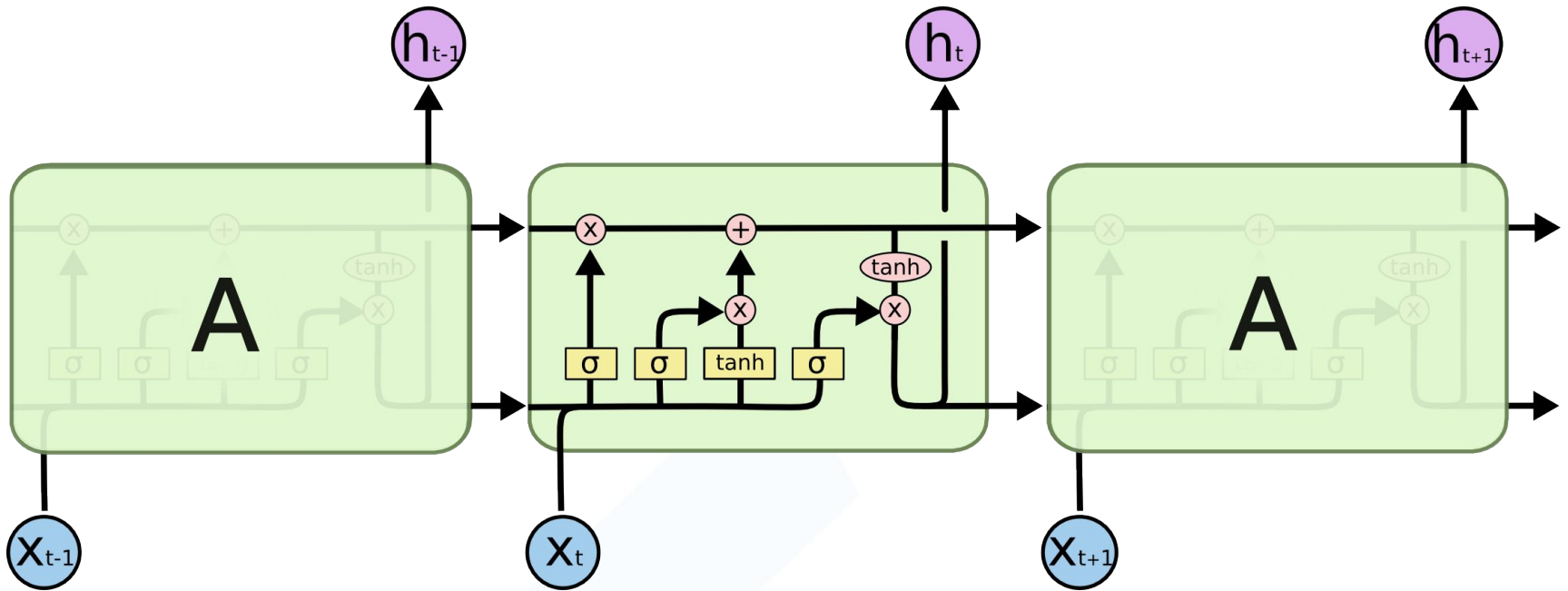
Encapsulated

Isolated

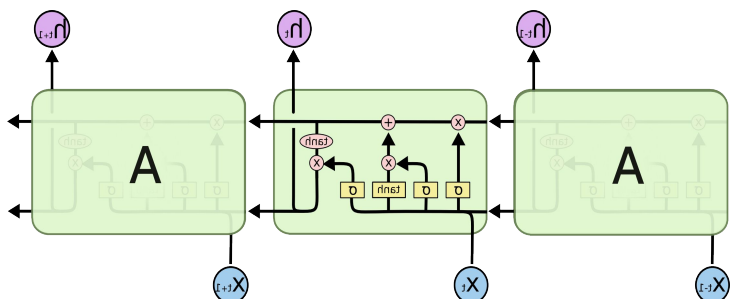
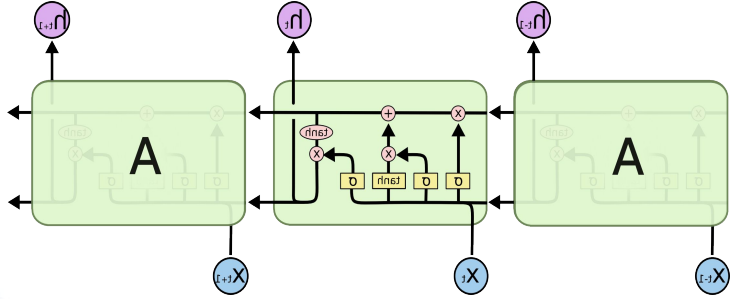
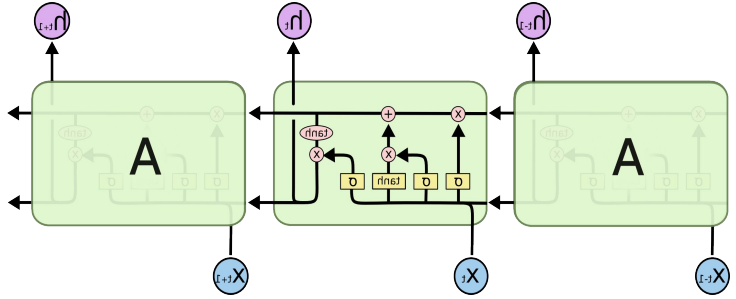
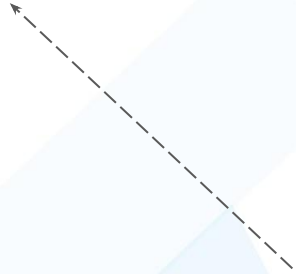
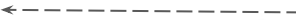
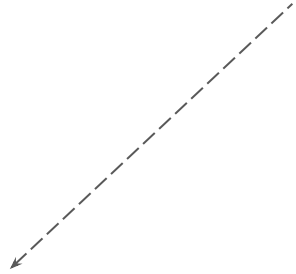
Generic

Composable

Order item micromodel



D
E
N
S
E



ciroc vodka
moet champagne

0.997286

fanta
doner kebab

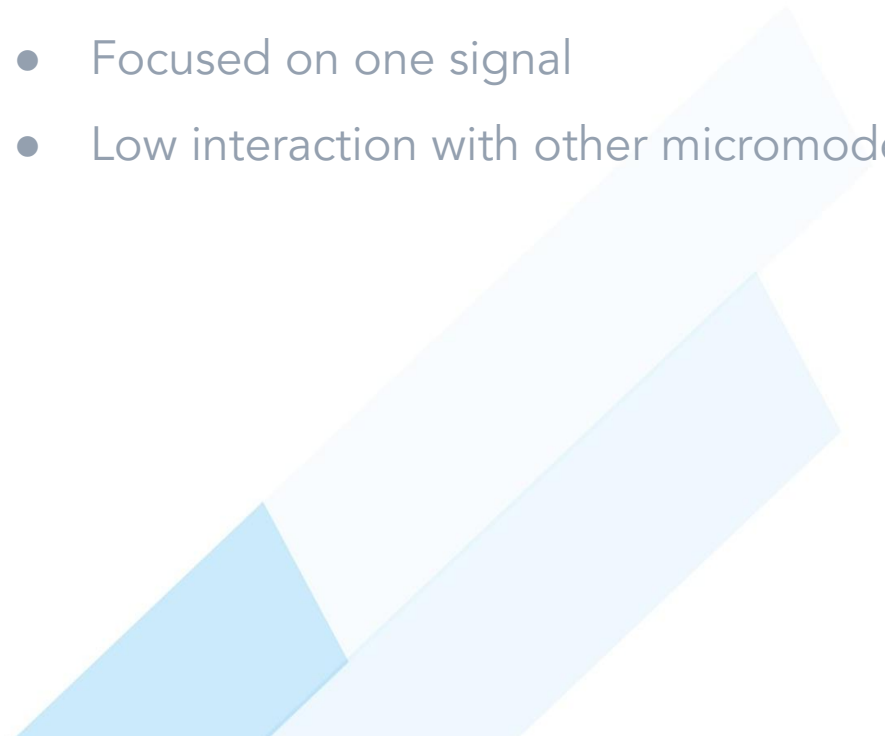
0.599078

mango sea-bass
kale and tofu

0.097097



Why is the order item model a good micro-model

- Simple input
 - Generic
 - Focused on one signal
 - Low interaction with other micromodels
- 

We can do something similar with email address

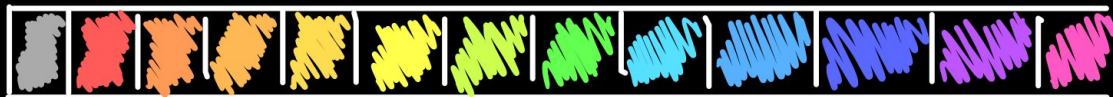
4934jf33jkf@yopmail.com

0.997286

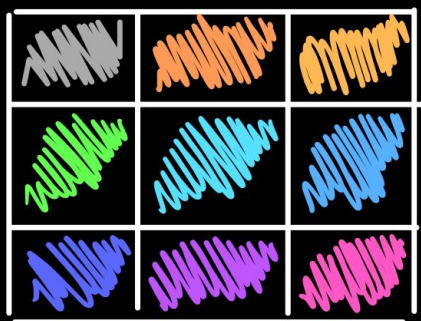
charles.xavier@university.edu

0.059302

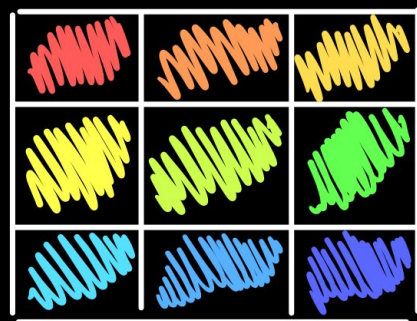




Model component library



Client A model



Client B model

Conclusion

- Deploy often
- Quantify what good looks like.
- Automate everything
- Simplicity over performance
- One code path everywhere



Thank you