



agraiculture

transforming the way farmers do commerce.



PROBLEM 1

smallholder farmers

do not know the prices of
their crops in real time, let
alone future prices.



PROBLEM 2

smallholder farmers

wait for prices to increase,
without understanding that
prices can decrease too.



PROBLEM 3

smallholder farmers

on average, earn \$2 a day.



TECH & FARMING

opportunity

25% of farmers in developing countries have smartphones.



SOLUTION

intuitive platform

that displays real-time and forecasted commodity prices, so farmers know when to sell their crops at optimal prices.



OUR **purpose**

our purpose is to determine the
direction/trend of the markets accurately, not
snipe exact prices.



agraiculture
platform

1

CROP selection

1. coffee
2. cocoa
3. corn

2

TIME duration

up to 7-day forecast to
avoid crops from
perishing.

3

FORECAST projection

farmers will know when to sell
their crops optimally.
comparing real-time prices
with forecasted prices.



FORECASTING USING ML ALGORITHMS: NEURAL NETWORKS

applying multiple (statistical) models

1. Mini-Rocket
2. Auto-Regressive Integrated Moving Average (ARIMA)
3. Gradient Boosted Machines (LightGBM (fig.1))
4. Recurrent Neural Networks (RNN (fig.2)/LSTM)
 - Multiple Views (weather, CFTC)
 - Avoiding Catastrophic Forgetting

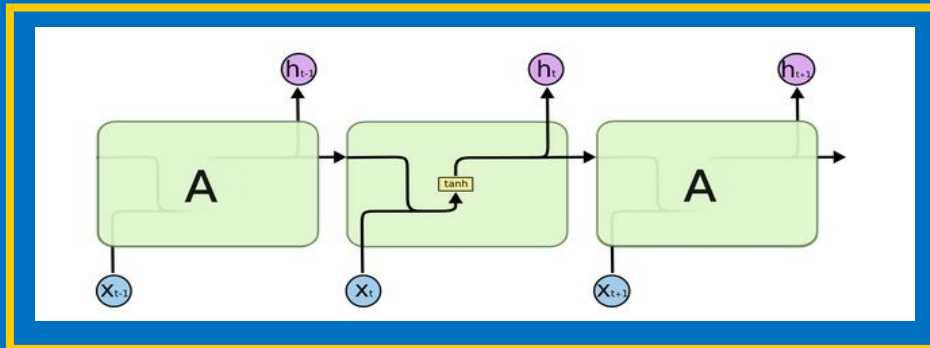


fig. 2: Recurrent Neural Network Structure. Implicitly encodes structure.

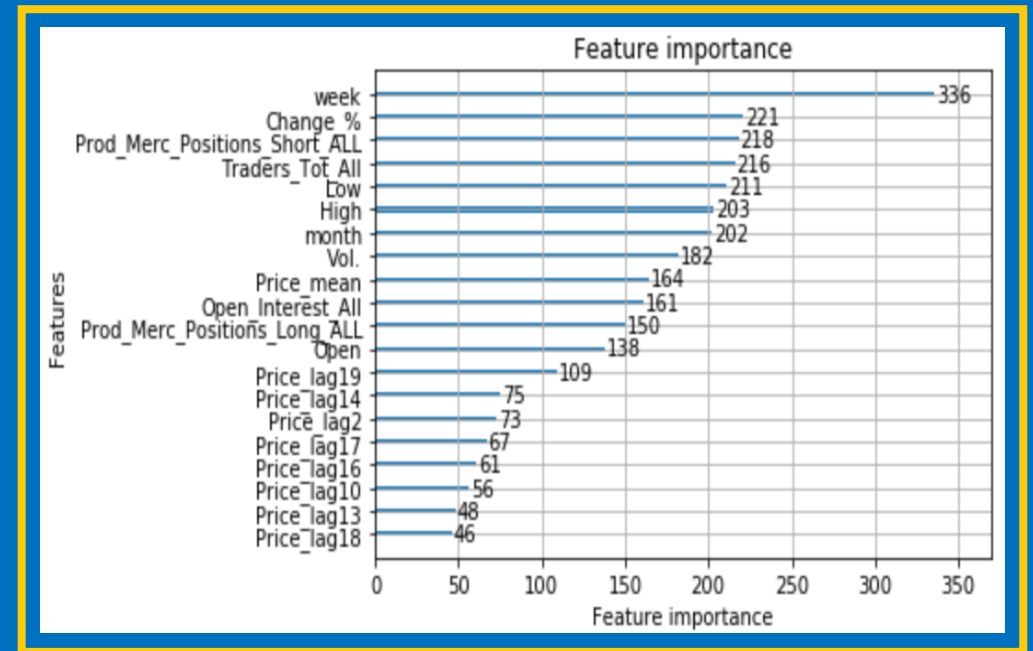


fig. 1: Feature Importances from statistical models.



data & model

coffee (C), cocoa (CC), corn (CZ)

- data we imported from:
 - Investing.com
 - CFTC
 - Weather Data (major parts in Brazil)
- trained coffee (C) for 6000 years (375 epochs)
- trained cocoa (CC) for 8000 years (500 epochs)
- corn (CZ) model is being developed



Investing.com





performance

coffee : (C), cocoa : (CC)

- model trained under 6000 years of the coffee (C) market and 8000 years of the cocoa (CC) market
- takes 250 days in the past to predict the following day
- according to figure 1 and 2, the **green line** is the true data and the **red line** is our forecasted price; they are nearly aligned
- our price predictions have measured to +/- \$7 on average for each prediction
- On some cases, price predictions are +/- \$0.02 off

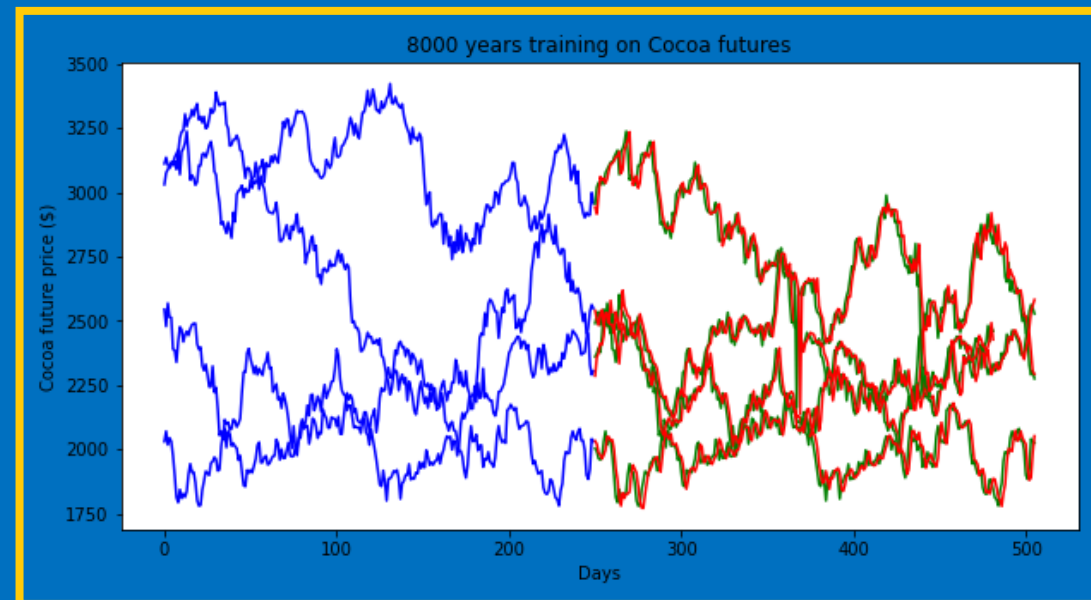


fig. 1 — trained data — truth — forecast

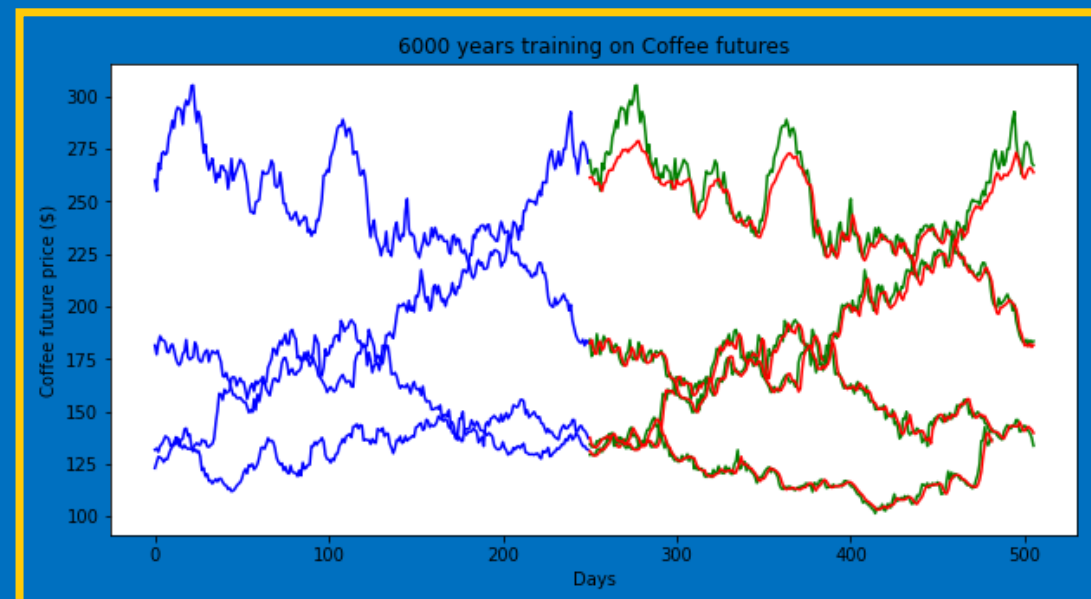


fig. 2 — trained data — truth — forecast



pilot program

- conducting a 1 month pilot program with a cooperative in Perú
- membership size of 200 – 300 farmers
- the cooperative sells 6,000 metric tons of coffee annually ~ 24 – 30 containers

cooperatives

- have larger yields
- stronger bidding power due to membership size
- formal banking



business model

SaaS
feature based
pricing model



cooperatives



3 pricing
models



standard



silver



gold



our competitive advantage

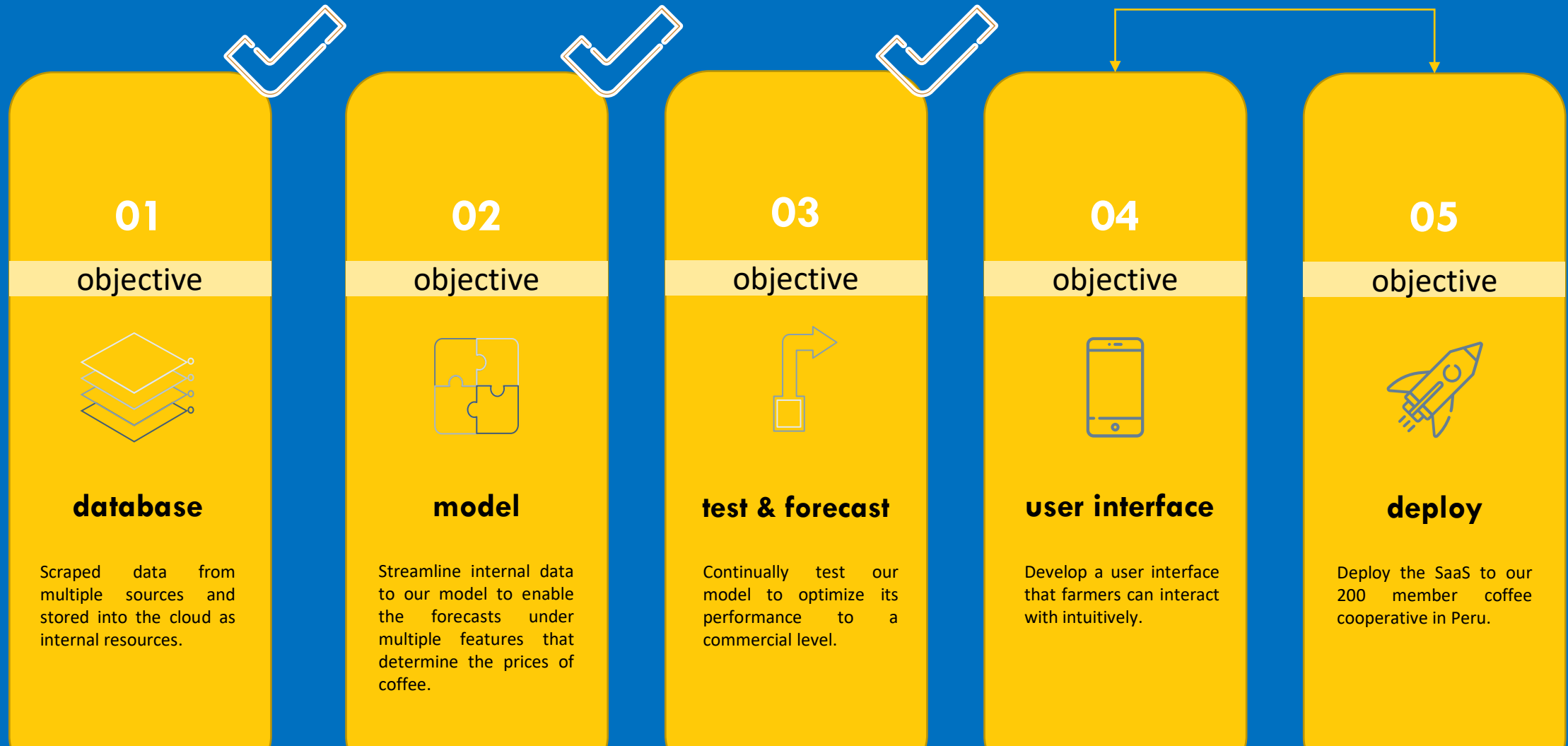
- we are pioneers in the space of forecasting commodity prices for smallholder farmers
- established relationships with cooperatives already built in Perú and Honduras
- strategic partnerships that will allow us to have immediate customers
- user interface is simple and intuitive to adopt

timing is important

in the past, smartphones were not prevalent in developing countries, even more so with farmers.
now, 25% of farmers in developing countries have smartphones.

completed objectives & timeline

2 MONTHS





partnerships & support





team



joshua simangunsong
ex-inmate @ jpm
returned peace corps volunteer
consultant @ deloitte



bijan varjavand
machine learning dude
AI research engineer



victor ganoza
former country director of technoserve
phd econ & mba



our ask

\$5,000

1. spanish speaking customer service team
2. freight tracker API

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