



Timing is Important: Risk-aware Fund Allocation based on Time-Series Forecasting

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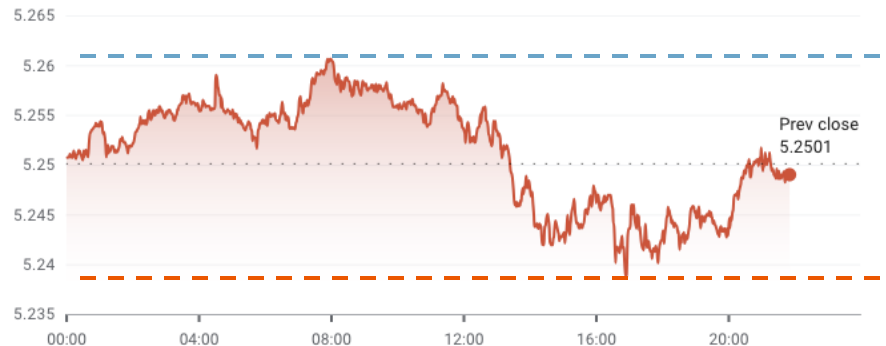
When shall I buy CAD

Canadian Dollar to Chinese Yuan

5.2490 ↓ 0.021% -0.0011 Today

8 Jul, 21:51:24 UTC · Disclaimer

1D 5D 1M 6M YTD 1Y 5Y MAX



- Fuyuan plans to attend the KDD
- He needs to buy 100 CAD cash
- When is the optimal timing?

Max: 5.261

$$\text{Diff} = 100 * (5.261 - 5.239) = 2.2 \text{ CNY}$$

Min: 5.239


≈ 0.5 * 

Image Captured from Google Finance

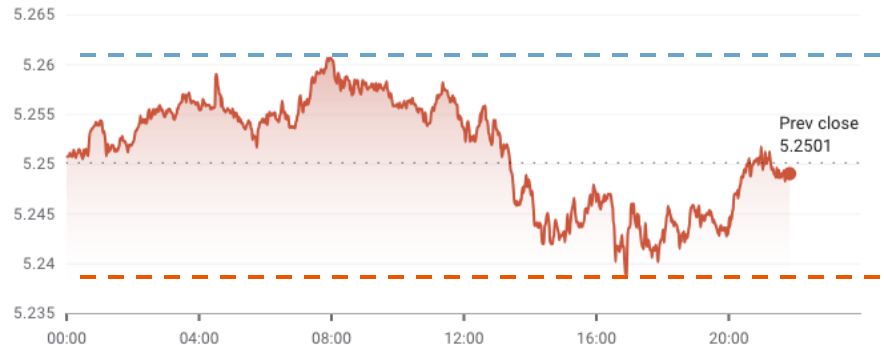
When shall Tencent buy CAD

Canadian Dollar to Chinese Yuan

5.2490 ↓ 0.021% -0.0011 Today

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1D 5D 1M 6M YTD 1Y 5Y MAX



As the volume increase, the gap becomes non-neglectable

Max: 5.261

$$\text{Diff} = 100\text{M} * (5.261 - 5.239) = 2.2 \text{ M CNY}$$

Min: 5.239

$$\approx 0.5\text{M} * \text{🍷}$$

Image Captured from Google Finance



Business Background

Canadian Dollar to Chinese Yuan

5.2490

↓ 0.021% -0.0011 Today

8 Jul, 21:51:24 UTC · Disclaimer

1D 5D 1M 6M YTD 1Y 5Y MAX

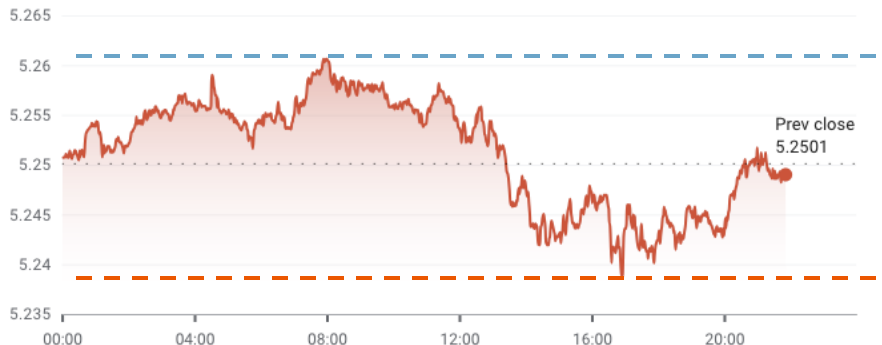
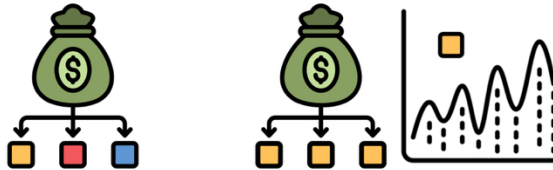


Image Captured from Google Finance

- WeChat Pay Oversea is 24/7 available.
- But Currency Market is not.
- This requires WeChat to serve as an exchanger sometime.
- The exchange volume on weekend is particularly high!
- Need to buy huge cash reserve before Friday.

Predict-then-Optimize



F. A. over Assets F. A. over Time

*Definition: Buy certain amount of assets
over a period of time, while its price varies.*

E.g. Buy 1k USD over Friday

Predict-then-Optimize



F. A. over Assets



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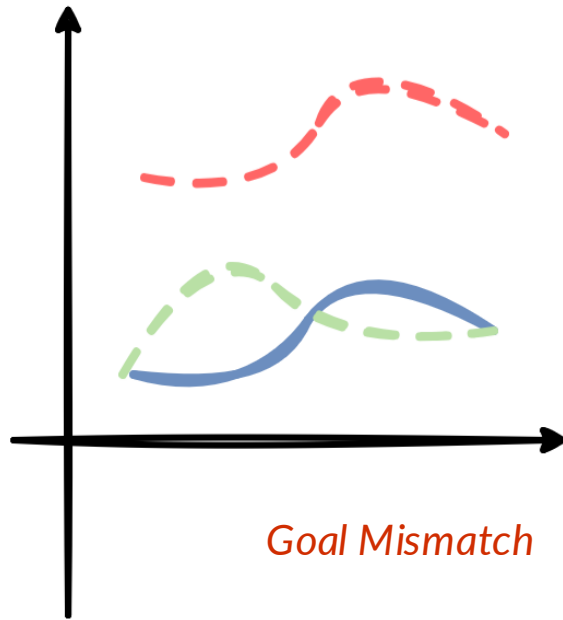
E.g. Buy 1k USD over Friday

A Predict-then-Optimize (PtO) framework:

- A forecasting model: $y_T = M(x_T)$
- An allocation model: $\min_a a * y_T, \text{ s.t. } \sum a = 1$



Why not two-stage



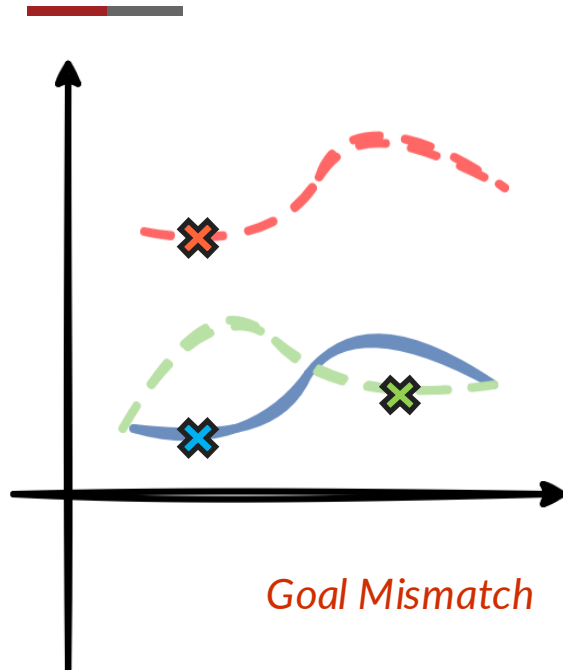
Prediction: Green > Red

Optimization: Red > Green

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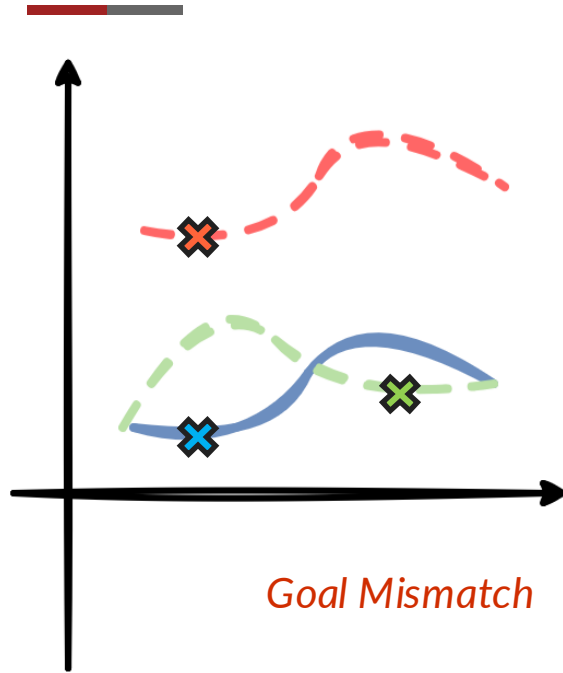
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Why not two-stage



Require a new training framework to “backprop” optimization feedback into training of prediction

Prediction: Green > Red

Optimization: Red > Green

A Predict-then-Optimize (PtO) framework:

- A forecasting model: $y_T = M(x_T)$
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RTS-PnO



- Goal Mismatch between prediction and allocation stages



RTS-PnO

- Goal Mismatch between prediction and allocation stages
- Apply the SPO+[1] loss to time series for goal alignment

$$L_o = \frac{1}{|D|} \min_{M(\cdot)} \sum_D l_o(a^*(\hat{y}_T), a(y_T))$$

$$l_o(a^*(\hat{y}_T), a(y_T)) \triangleq 2a^*(y_T)\hat{y}_T - a^*(y_T)y_T + \max_{a \in A} \{ay_T - 2a\hat{y}_T\}$$

Uncertainty of Forecasting



- The uncertainty of forecasting varies across steps



Uncertainty of Forecasting



Algorithm 1 Calculating Positional Uncertainty for Forecasting Model

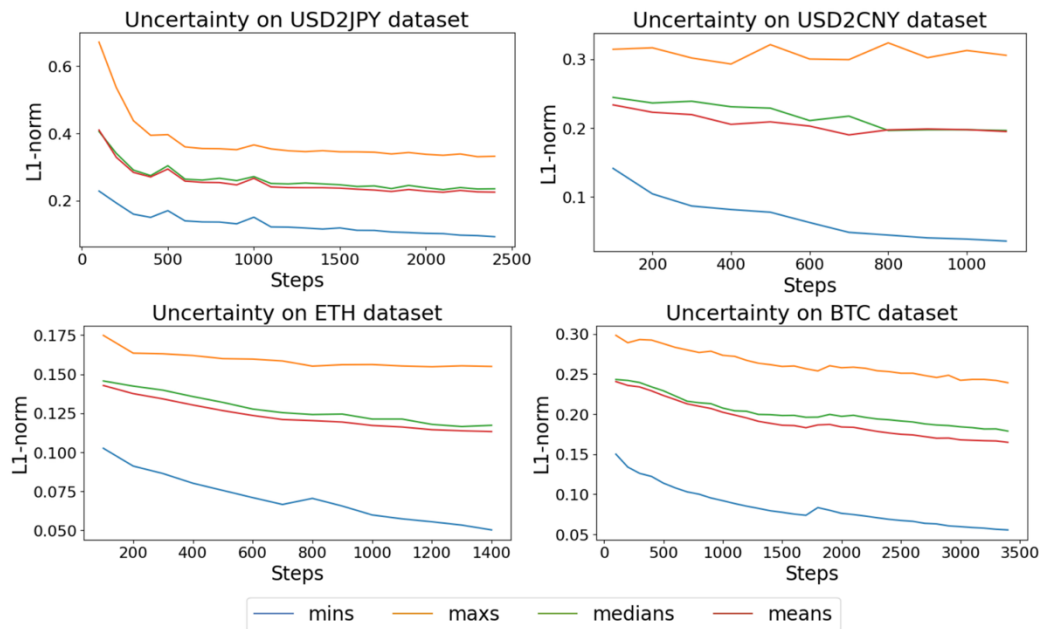
Require: Calibration Dataset \mathcal{D}_c , coverage rate γ

Ensure: Positional Uncertainty \mathbf{r}

- 1: Initialize Positional Uncertainty Sets $\epsilon_1 = \{ \}, \dots, \epsilon_H = \{ \}$
 - 2: **for** for data instance (x_T, y_T, c_T) in Calibration Set \mathcal{D}_c **do**
 - 3: Calculate $\hat{y}_T = [\hat{p}_{T+1}, \dots, \hat{p}_{T+H}]$ given Eq. 2
 - 4: **for** h in $1, \dots, H$ **do**
 - 5: $\epsilon_h \leftarrow \epsilon_h \cup \{|\hat{p}_{T+h} - p_{T+h}|\}$
 - 6: **for** h in $1, \dots, H$ **do**
 - 7: $r_h = \left(\frac{|\mathcal{D}_c|+1}{|\mathcal{D}_c|} \gamma \right)$ - quantile in ϵ_h
 - 8: **Return** $\mathbf{r} = [r_1, r_2, \dots, r_H]$
-

- The uncertainty of forecasting varies across steps
- Positional-aware risk on forecasting

Why not a fixed threshold



The threshold varies during training phase

Evaluation

Category	Dataset	Forecasting-Only				Risk-Avoid				RTS-PtO		RTS-PnO		Relative Improvement	
		Top-1		Top-5		Top-1		Top-5							
		regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret(%)	R.R.(%)
Currency	USD2CNY	36.88	5.10	37.00	5.12	35.80	4.95	35.83	4.96	<u>35.74</u>	<u>4.94</u>	31.68	4.38	12.82%	12.79%
	USD2JPY	54.50	34.92	54.21	34.73	<u>49.66</u>	<u>31.90</u>	50.01	32.12	52.11	32.66	48.77	31.25	1.82%	2.08%
	AUD2USD	19.56	29.60	19.92	30.15	<u>19.38</u>	<u>29.36</u>	19.49	29.52	19.48	29.51	19.06	28.84	1.68%	1.80%
	NZD2USD	17.43	28.75	17.66	29.14	<u>16.54</u>	<u>27.29</u>	16.64	27.44	16.82	27.75	15.68	25.85	5.48%	5.57%
Stock	S&P 500	134.99	4.25	135.47	4.24	122.50	3.84	124.24	<u>3.90</u>	126.06	3.94	<u>124.05</u>	<u>3.90</u>	-1.27%	-1.56%
	Dow Jones	1090.88	4.16	1075.79	4.09	<u>1022.73</u>	<u>3.91</u>	1032.21	3.93	1022.90	3.92	997.52	3.82	2.53%	2.36%
Cryptos	BTC	2159.78	4.46	2167.96	4.47	<u>1856.21</u>	<u>3.90</u>	1858.57	3.91	1924.65	3.96	1843.26	3.70	0.70%	5.41%
	ETH	151.14	5.56	149.61	5.48	<u>131.41</u>	4.68	131.42	4.68	138.60	4.96	131.40	<u>4.73</u>	0.00%	-1.07%
Avg. Rank		5.38	5.5	5.63	5.5	2	1.88	3.38	3.13	3.5	3.5	1.13	1.25		

Setup:

- Datasets: 4 * Currency + 2 * Stock + 2 * Cryptos
- Backbone: PatchTST

$$\text{regret} \triangleq |a^*(y_T) \cdot y_T - a^*(\hat{y}_T) \cdot y_T|$$

$$\text{R. R.} \triangleq \frac{\text{regret}}{|a^*(y_T) \cdot y_T|}$$

Evaluation

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		regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret(%)	R.R.(%)
Currency	USD2CNY	36.88	5.10	37.00	5.12	35.80	4.95	35.83	4.96	<u>35.74</u>	<u>4.94</u>	31.68	4.38	12.82%	12.79%
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Avg. Rank		5.38	5.5	5.63	5.5	2	1.88	3.38	3.13	3.5	3.5	1.13	1.25		

- RTS-PnO proves to be effective:
- Forecasting-Only not reliable:
- Risk-Avoid is effective:

RTS-PnO > others

others > Forecasting-Only

Good on Stock & Crypto

Evaluation

Forecasting Model	Dataset	Forecasting-Only				Risk-Avoiding				RTS-PtO		RTS-PnO		Relative Improvement	
		Top-1		Top-5		Top-1		Top-5							
		regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓	regret(%)	R.R.(%)
DLinear	USD2CNY	36.99	5.12	36.73	5.08	35.50	<u>4.91</u>	38.11	5.27	<u>35.31</u>	4.98	34.88	4.81	1.23%	3.50%
	Dow Jones	1103.11	4.21	1128.71	4.24	1036.65	3.96	1075.97	4.08	1073.30	4.10	<u>1042.35</u>	<u>3.98</u>	-0.55%	-0.51%
TimesNet	USD2CNY	39.77	5.50	39.46	5.46	36.83	5.09	37.47	5.18	<u>35.99</u>	<u>4.98</u>	33.73	4.66	6.70%	6.87%
	Dow Jones	1157.76	4.40	1143.82	4.32	<u>1037.71</u>	3.98	1082.45	4.11	1042.67	<u>3.95</u>	972.51	3.74	6.70%	5.61%
FEDFormer	USD2CNY	36.44	5.04	36.89	5.10	36.28	5.02	36.53	5.05	<u>35.94</u>	<u>4.97</u>	32.32	4.47	11.23%	11.19%
	Dow Jones	1087.49	4.15	1100.99	4.19	1065.08	4.05	1078.61	4.09	<u>1043.41</u>	<u>3.98</u>	1010.96	3.82	3.21%	4.19%

- RTS-PnO is model-agnostic:

Good on other TS models

Ablation

Dataset	PtO		Fixed-PnO		Adaptive-PnO	
	regret↓	R.R.↓	regret↓	R.R.↓	regret↓	R.R.↓
USD2CNY	35.74	4.94	<u>34.66</u>	<u>4.71</u>	31.68	4.38
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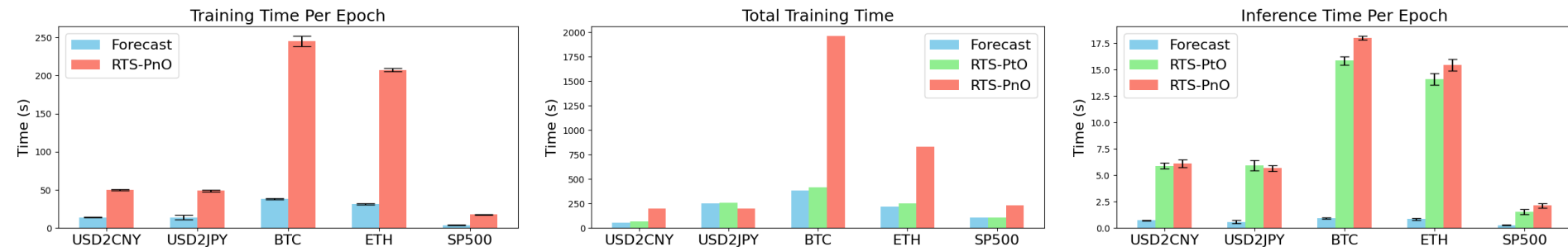
Category	Dataset	Prediction		RTS-PnO	
		MSE	MAE	MSE	MAE
Currency	USD2CNY	0.0049	0.0397	<u>0.0053</u>	<u>0.0430</u>
	USD2JPY	0.0383	0.1263	<u>0.1201</u>	<u>0.2796</u>
	AUD2USD	0.0277	0.1220	<u>0.0350</u>	<u>0.1439</u>
	NZD2USD	0.0233	0.1072	<u>0.0327</u>	<u>0.1334</u>
Stock	S&P 500	0.1533	0.2744	<u>0.5567</u>	<u>0.6194</u>
	Dow Jones	0.1184	0.2354	<u>0.3552</u>	<u>0.4815</u>
Criptos	BTC	0.0197	0.0962	<u>0.0953</u>	<u>0.2321</u>
	ETH	0.0213	0.1003	<u>0.1297</u>	<u>0.2608</u>

- Ablation on Risk Threshold:
- Ablation on Prediction Performance:

Adaptive > Fixed

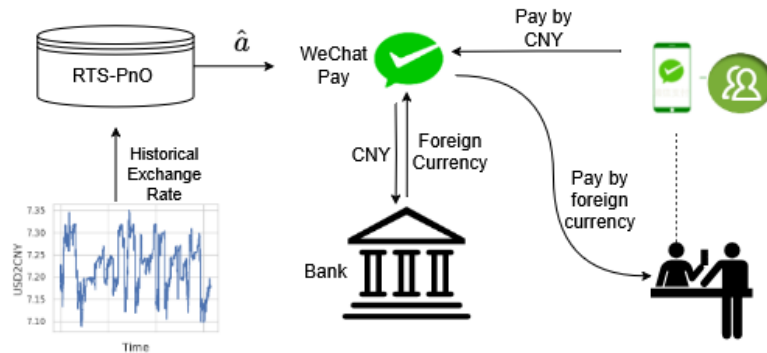
Sacrifice the prediction accuracy

Efficiency Analysis

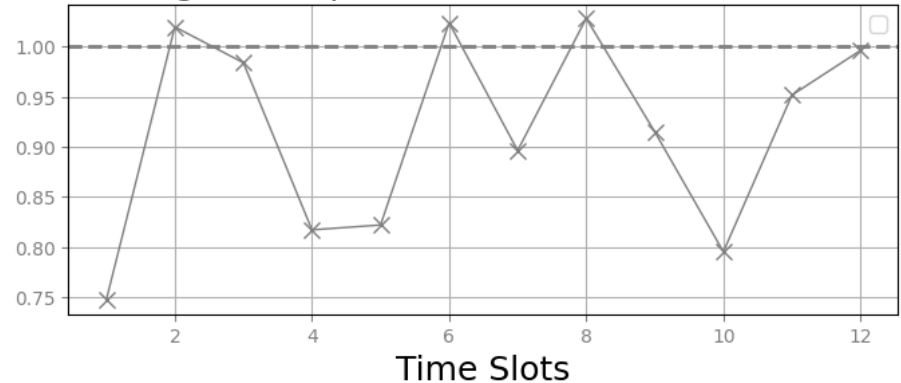


- Increase on both training and inference time
- Inference increase due to the additional allocation stage
- Training increase due to the additional proxy loss

Online Experiment



Regret Compared to Product-line Baseline



- Tencent requires to provide oversea services to customers during weekend
- A 8.4% decrease in terms of relative regret



Summary

1. Asset Allocation over Time
2. RTS-PnO:
 - Predict-and-Optimize framework
 - Adaptive risk-aware allocation
3. Evaluation on offline Financial AI datasets and online experiment



Thanks for listening!

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2. RTS-PnO:
 - Predict-and-Optimize framework
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Code available at <https://github.com/fuyuanlyu/RTS-PnO>