

Peramalan Permintaan Oli Dengan Integrasi Support Vector Regression (SVR) Dan Particle Swarm Optimization (PSO)

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Pendahuluan

Gudang Spare Part dan Consumable Part



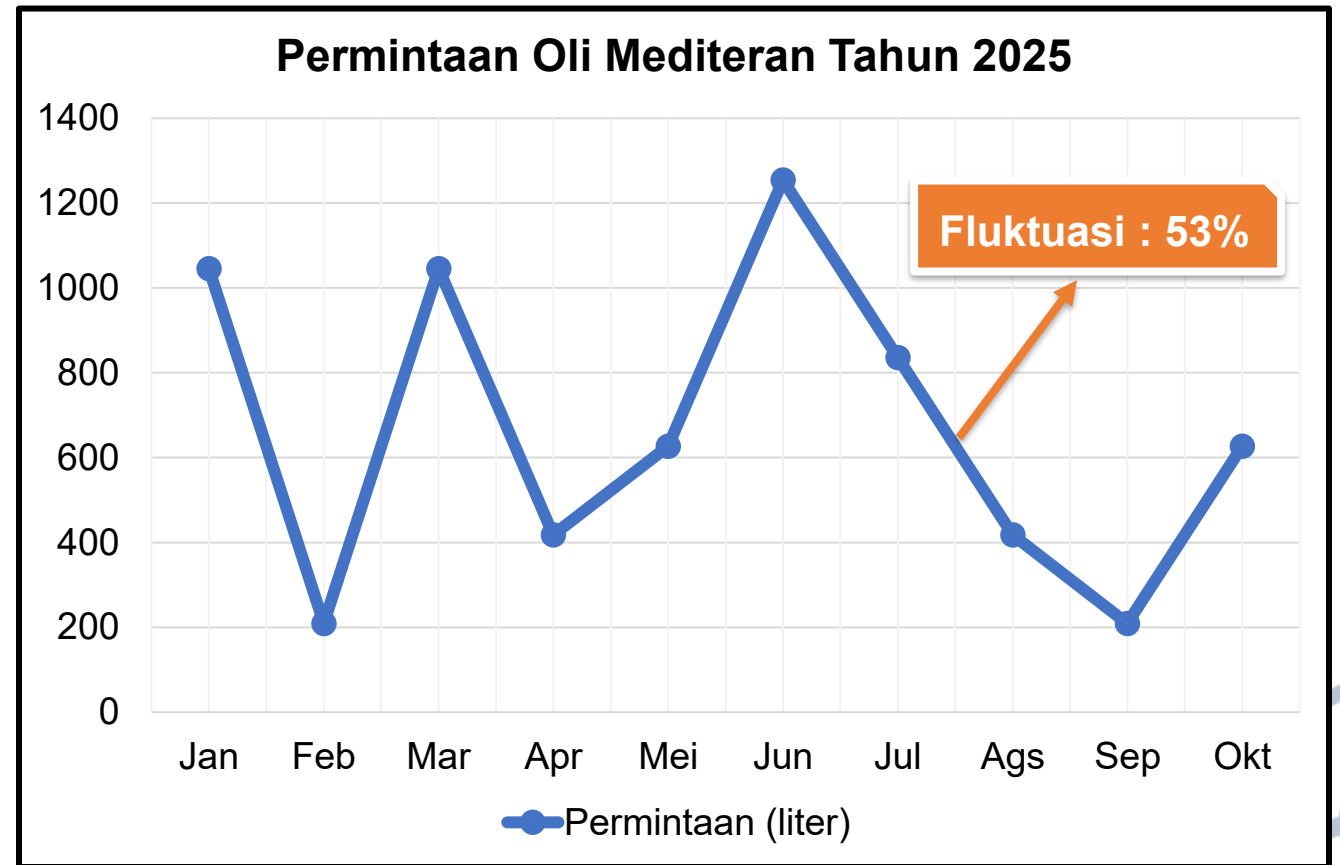
Belum Diterapkan Sistem
Stok Aman (*Safety Stock*) Pada Oli



Alasan : Jumlah permintaan oli tidak
menentu (fluktuatif)



Solusi : Peramalan (*forecasting*)

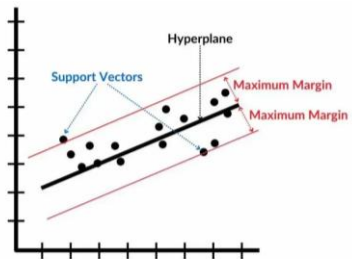


Rumusan Masalah

- ❑ Berapa **kuantitas** permintaan oli pada periode berikutnya?
- ❑ Bagaimana cara menghasilkan peramalan dengan tingkat **kesalahan (*error*) minimum?**

Metode

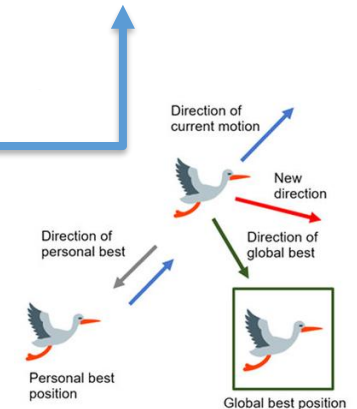
Membangun **model prediksi** dari data non-linier



**Support
Vector
Regression
(SVR)**

**Particle
Swarm
Optimization
(PSO)**

Mencari **parameter terbaik** agar SVR bekerja dengan optimal



Pembahasan

Inisialisasi Partikel Awal

X	C	ε	σ	cLR	λ	F1	F2	F3
1	328,3279	0,0399	5,5198	1,7243	0,6502	1	1	0
2	961,2476	0,0133	2,0424	3,9132	0,3644	0	1	1
3	849,0079	0,0564	2,2829	2,2771	1,5674	1	1	0

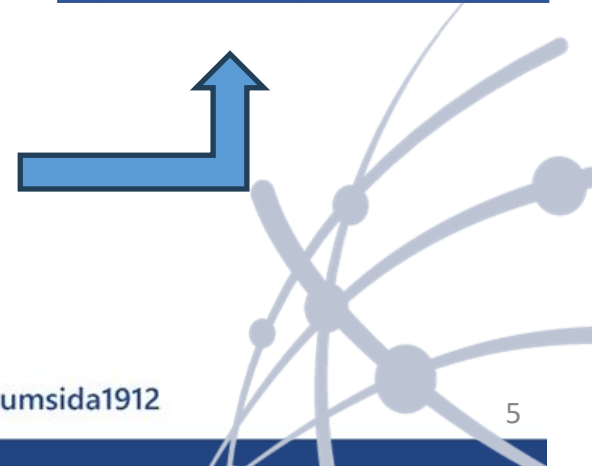
Hasil Prediksi Awal

Uji	F(x)	Denormalisasi	Aktual
1	14,8774	2549,144	159
2	14,9088	2554,226	206
3	14,9008	2552,929	193
4	14,9043	2553,500	258
5	14,9235	2556,603	168

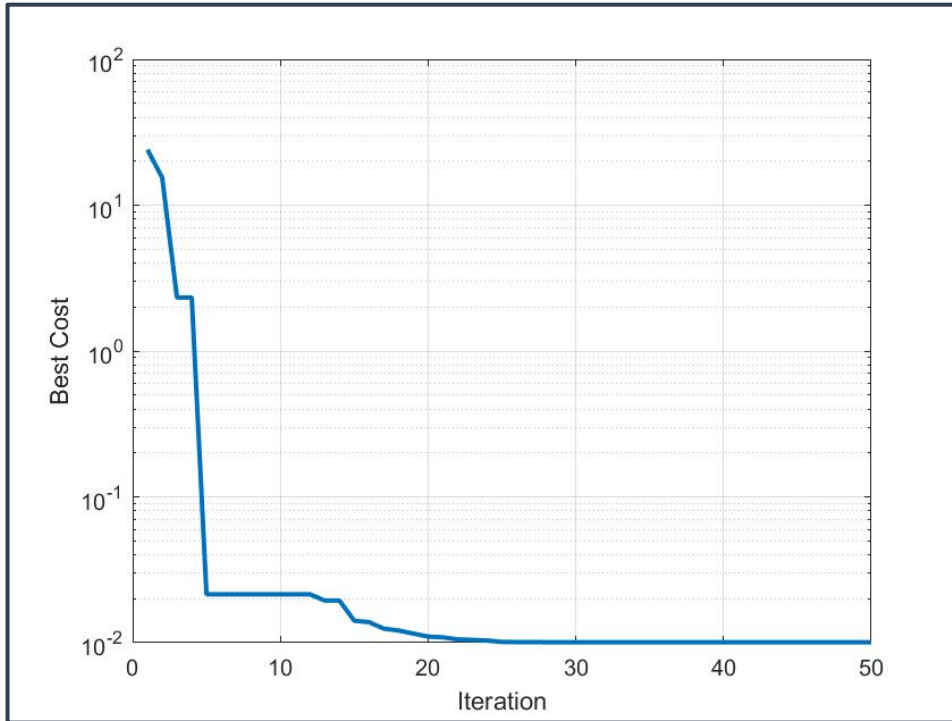
Nilai Error Awal

Fold	Error
1	11,77
2	10,07
3	11,14
4	11,92
5	9,60
Cost	10,90

Nilai *error* tinggi:
Perlu dilakukan
optimasi parameter



Pembahasan



Grafik Optimasi PSO-MATLAB

Parameter Optimal PSO

X	C	ε	σ	cLR	λ
1	199,9993	0,0744	0,2487	0,1954	0,003

Nilai Cost Setelah Optimasi

gBest	Cost
1	24,03988
2	15,42977
3	2,32748
4	2,32748
5	0,02146
...	...
50	0,01006

Selanjutnya parameter yang optimal ini **dapat digunakan** pada proses prediksi pada tahap berikutnya



Error Kecil

Pembahasan

FOLD K = 1

Data Uji (Data *Testing*)

Indeks	F1	F2	F3	Aktual
0	0,9630	0,2099	0,2840	0,1235
1	0,2099	0,2840	0,1235	0,4136
2	0,2840	0,1235	0,4136	0,3333
...	...	...	...	...
9	0,9815	0,3519	0,7716	0,2222

Data Latih (Data *Training*)

Indeks	F1	F2	F3	Aktual
10	0,3519	0,7716	0,2222	0,0741
11	0,7716	0,2222	0,0741	0,2963
12	0,2222	0,0741	0,2963	1,0000
...	...	...	...	...
24	0,3580	0,6914	0,6667	0,3210

Matriks Kernel *Radial Basis Function* (RBF)

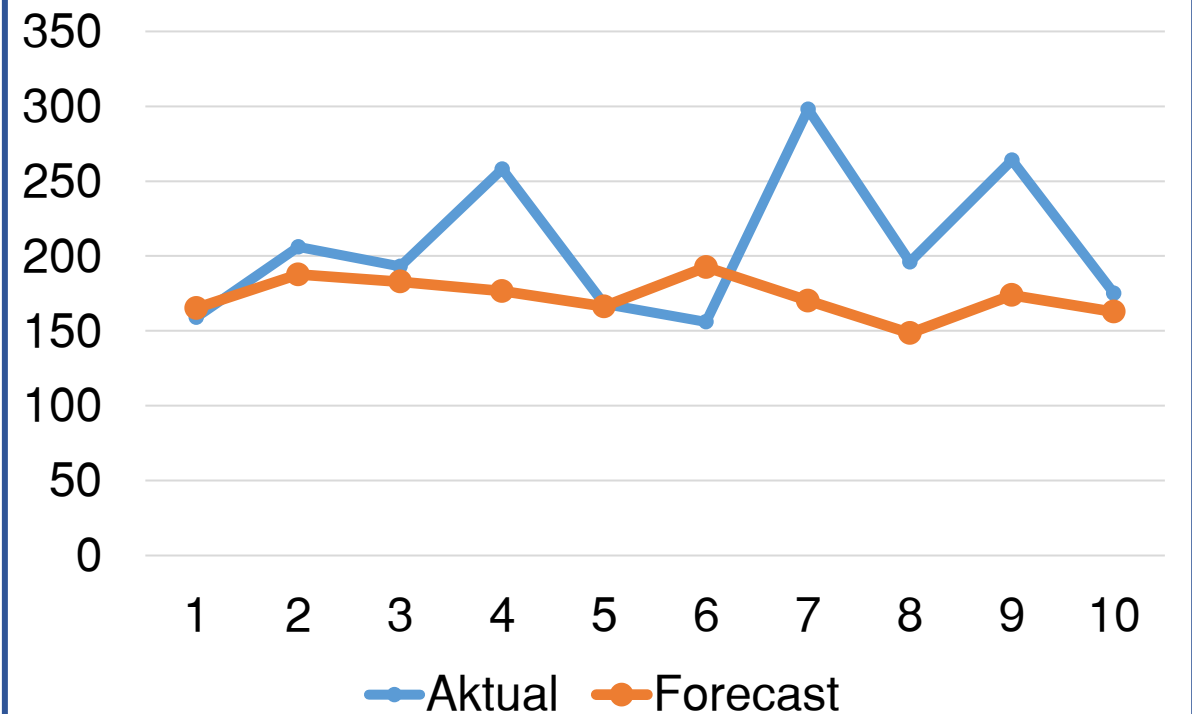
Uji/Latih	1	2	3	...	15
1	0,0474	0,5210	0,0118	...	0,0159
2	0,7852	0,0765	0,7845	...	0,0771
3	0,7166	0,0576	0,8676	...	0,5700
4	0,5936	0,0195	0,9140	...	0,2611
5	0,1162	0,0104	0,1574	...	0,9397
6	0,9823	0,1936	0,8098	...	0,1455
7	0,2738	0,9814	0,0891	...	0,0248
8	0,0074	0,0001	0,0222	...	0,3464
9	0,5333	0,0148	0,8727	...	0,2674
10	0,0035	0,0137	0,0015	...	0,0395

Hasil

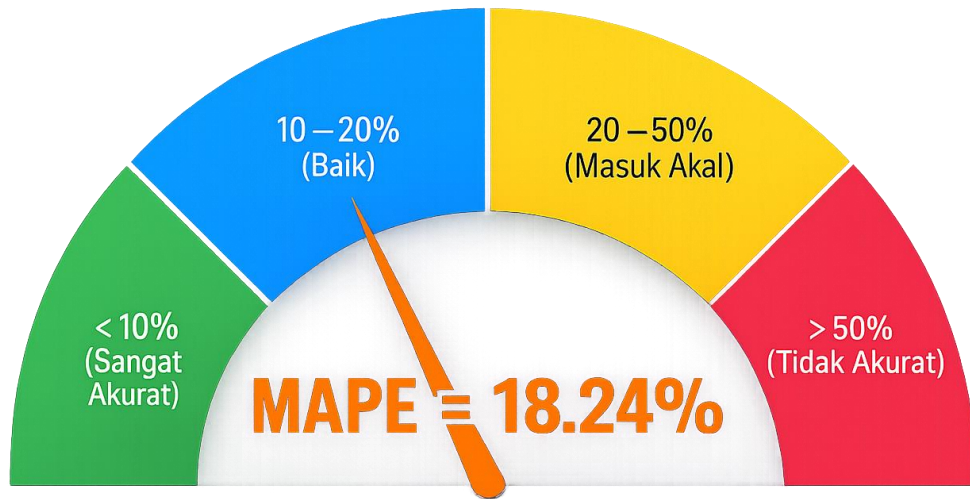
Tabel Hasil Peramalan

Uji	Aktual	Forecast
1	159	165,19
2	206	187,67
3	193	182,87
4	258	176,59
5	168	166,27
6	156	192,61
7	298	170,15
8	196	148,46
9	264	173,84
10	175	162,74

Grafik Hasil Peramalan



Temuan Penting Penelitian



Interpretasi Lewis (1982)

MAPE

(Mean Absolute Percentage Error)

- Nilai akurasi hasil peramalan

$$\begin{aligned}\text{MAPE} &= \left(\frac{100}{n}\right) \sum \left| A_t - \frac{F_t}{A_t} \right| \\ &= \left(\frac{100}{10}\right) \left[\left| 159 - \frac{165}{159} \right| + \dots + \left| 175 - \frac{163}{175} \right| \right] \\ &= \mathbf{18,24 \%}\end{aligned}$$

Manfaat Penelitian

Penelitian ini memberikan manfaat bagi perusahaan sebagai **dasar pertimbangan** dalam menentukan tingkat stok aman (*safety stock*) dan jumlah pemesanan optimal oli guna mendukung pengelolaan persediaan yang lebih efektif dan efisien

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