

# CBEM PROJECT 2

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# Project 2: Regression Analysis

- example is the model 1  $k_f = g(\mathbf{x}, T, \varphi, \dot{\varphi}) = x_6 \cdot e^{x_1 \cdot T} \cdot \dot{\varphi}^{x_2 + x_5 \cdot T} \cdot \varphi^{x_3} \cdot e^{x_4 \cdot \varphi}.$
- example is the model 2  $k_f = x_{10} \cdot e^{x_1 \cdot T} \cdot \dot{\varphi}^{x_2 + x_5 \cdot T + x_6 \cdot \varphi} \cdot \varphi^{x_3 + x_7 \cdot T + x_8 \cdot \dot{\varphi}} \cdot e^{x_4 \cdot \varphi + x_9 \cdot \dot{\varphi}}.$

# Source of the data

- Three data sheets from moodle.
  - C15
  - C6o
  - 100Cr6
- Each will have different R-square values.
- The values and graphs used below are from the Dataset -100Cr6

# Model 1

$$k_f = g(\boldsymbol{x}, T, \varphi, \dot{\varphi}) = x_6 \cdot e^{x_1 \cdot T} \cdot \dot{\varphi}^{x_2 + x_5 \cdot T} \cdot \varphi^{x_3} \cdot e^{x_4 \cdot \varphi}$$

```

1 function [kf_calc,opt_x]=modell(filename);
2 filename = 'Cr6.mat';
3 dataset = importdata(filename);
4
5 %Loading Data from Dataset to variables
6 phidot=dataset(:,1);
7 phi=dataset(:,2);
8 Temp=dataset(:,3);
9 kf_data=dataset(:,4);
10
11 %Assigning the parameters
12 B=log(kf_data);
13 t1=Temp;
14 t2=log(phidot);
15 t3=log(phi);
16 t4=phi;
17 t5=Temp.*log(phidot);
18 t6=ones(size(dataset,1),1);
19
20 %Putting the parameters in the array
21 A=[t1 t2 t3 t4 t5 t6];
22
23
24 %Ax=B
25 %x = A\B
26 opt_x=A\B;
27
28 modl_x1=opt_x(1,1);
29 modl_x2=opt_x(2,1);
30 modl_x3=opt_x(3,1);
31 modl_x4=opt_x(4,1);
32 modl_x5=opt_x(5,1);
33 modl_x6=exp(opt_x(6,1));
34
35 %Our prediction of Kf
36 kf_calc=exp(A*opt_x);
37 opt_x
38 modl_x6
39 disp('R_square of model(1)');
40 rsquare(kf_data,kf_calc)

```

# Calculated values from model 1

- R-square = 0.8648
- $Ax=B$
- Compute  $x = A \backslash B$
- $x$  is a 6x1 matrix
- $X =$ 
  - 0.0013
  - 0.0623
  - 0.1520
  - 0.1681
  - 0.0001
  - 1275.3

## Model 2

$$k_f = x_{10} \cdot e^{x_1 \cdot T} \cdot \dot{\varphi}^{x_2 + x_5 \cdot T + x_6 \cdot \varphi} \cdot \varphi^{x_3 + x_7 \cdot T + x_8 \cdot \dot{\varphi}} \cdot e^{x_4 \cdot \varphi + x_9 \cdot \dot{\varphi}}$$

# Calculated values from model 2

- R-square = 0.9385
- Calculation same as in the model 1
- x is a 10x1 matrix
- X=

-0.0016

0.0820

0.2586

-0.0297

-0.0001

-0.0521

-0.0003

0.0008

0.0010

```
1 function [kf_calc,opt_x]=model2(filename);
2 filename = 'Cr6.mat';
3 dataset = importdata(filename);
4
5 phidot=dataset(:,1);
6 phi=dataset(:,2);
7 Temp=dataset(:,3);
8 kf_data=dataset(:,4);
9
10 B=log(kf_data);
11 t1=Temp;
12 t2=log(phidot);
13 t3=log(phi);
14 t4=phi;
15 t5=Temp.*log(phidot);
16 t6=phi.*log(phidot);
17 t7=Temp.*log(phi);
18 t8=phidot.*log(phi);
19 t9=phidot;
20 t10=ones(size(dataset,1),1);
21
22 A=[t1 t2 t3 t4 t5 t6 t7 t8 t9 t10];
23
24 %Ax=B
25 %x = A\B
26 opt_x = A\B;
27
28 mod2_x1 = opt_x(1,1);
29 mod2_x2 = opt_x(2,1);
30 mod2_x3 = opt_x(3,1);
31 mod2_x4 = opt_x(4,1);
32 mod2_x5 = opt_x(5,1);
33 mod2_x6 = opt_x(6,1);
34 mod2_x7 = opt_x(7,1);
35 mod2_x8 = opt_x(8,1);
36 mod2_x9 = opt_x(9,1);
37 mod2_x10 = exp(opt_x(10,1));
38
39 kf_calc=exp(A*opt_x);
40 opt_x
41 mod2_x10
42 display('R_square of model(2)');
43 rsquare(kf_data,kf_calc)
```

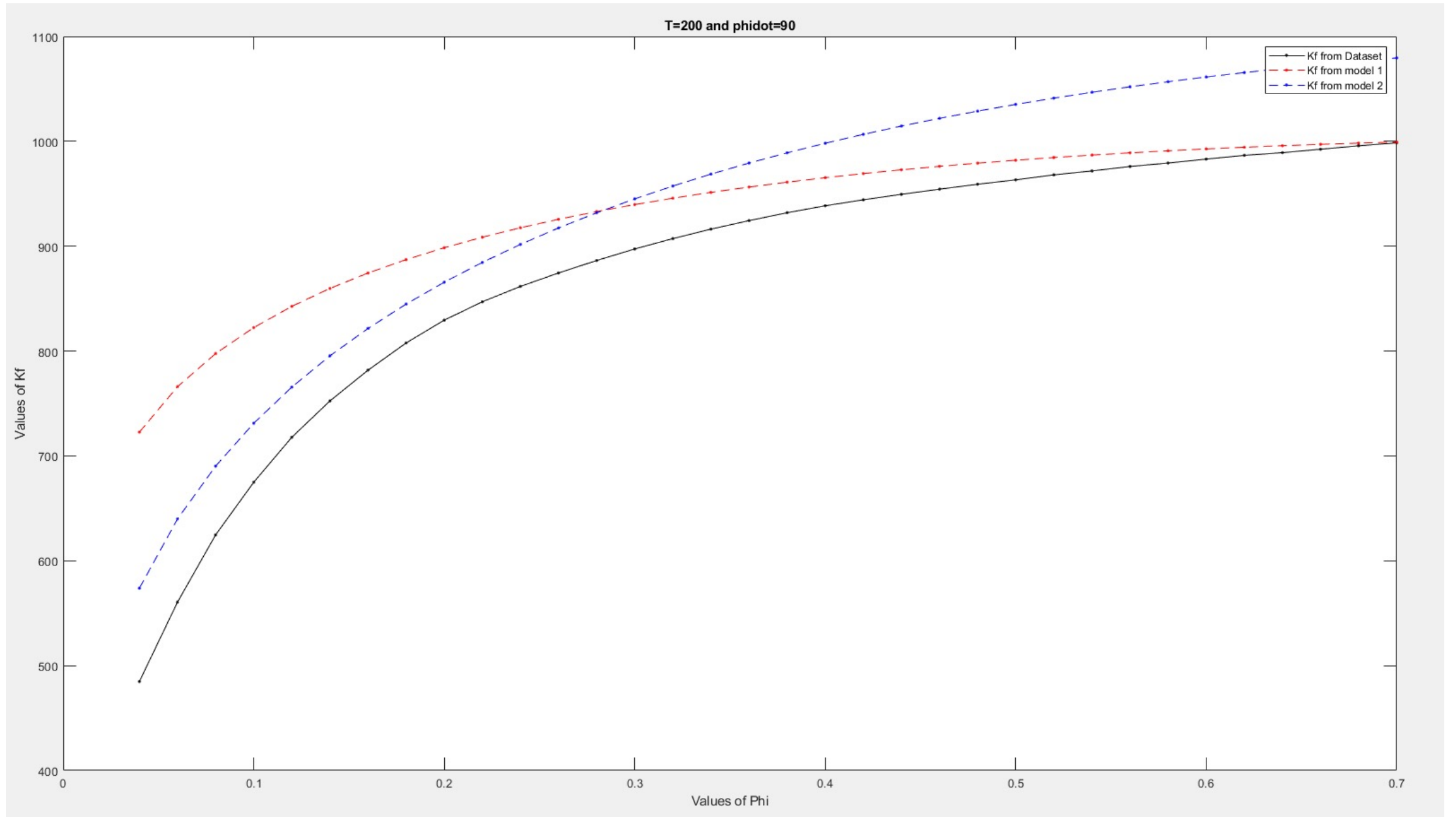
# R-squared

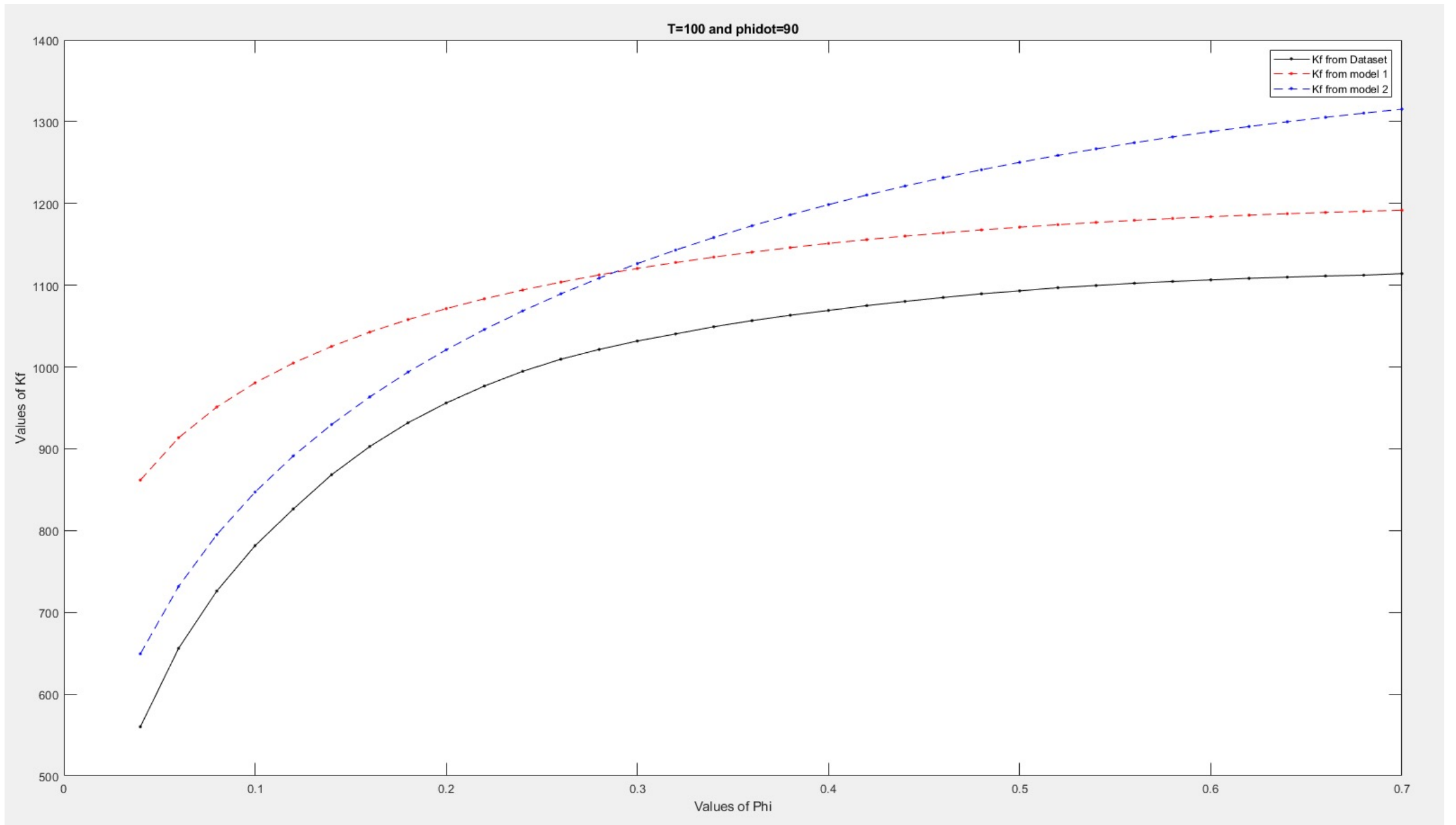
- the coefficient of determination for models 1 is  $R\text{-squared} = 0.8648$
- the coefficient of determination for models 2 is  $R\text{-squared} = 0.9385$



## 2 Dimensional graphs

We choose Temp and Phidot to be constant because they are the variables that change the least amount of times throughout the datasets.

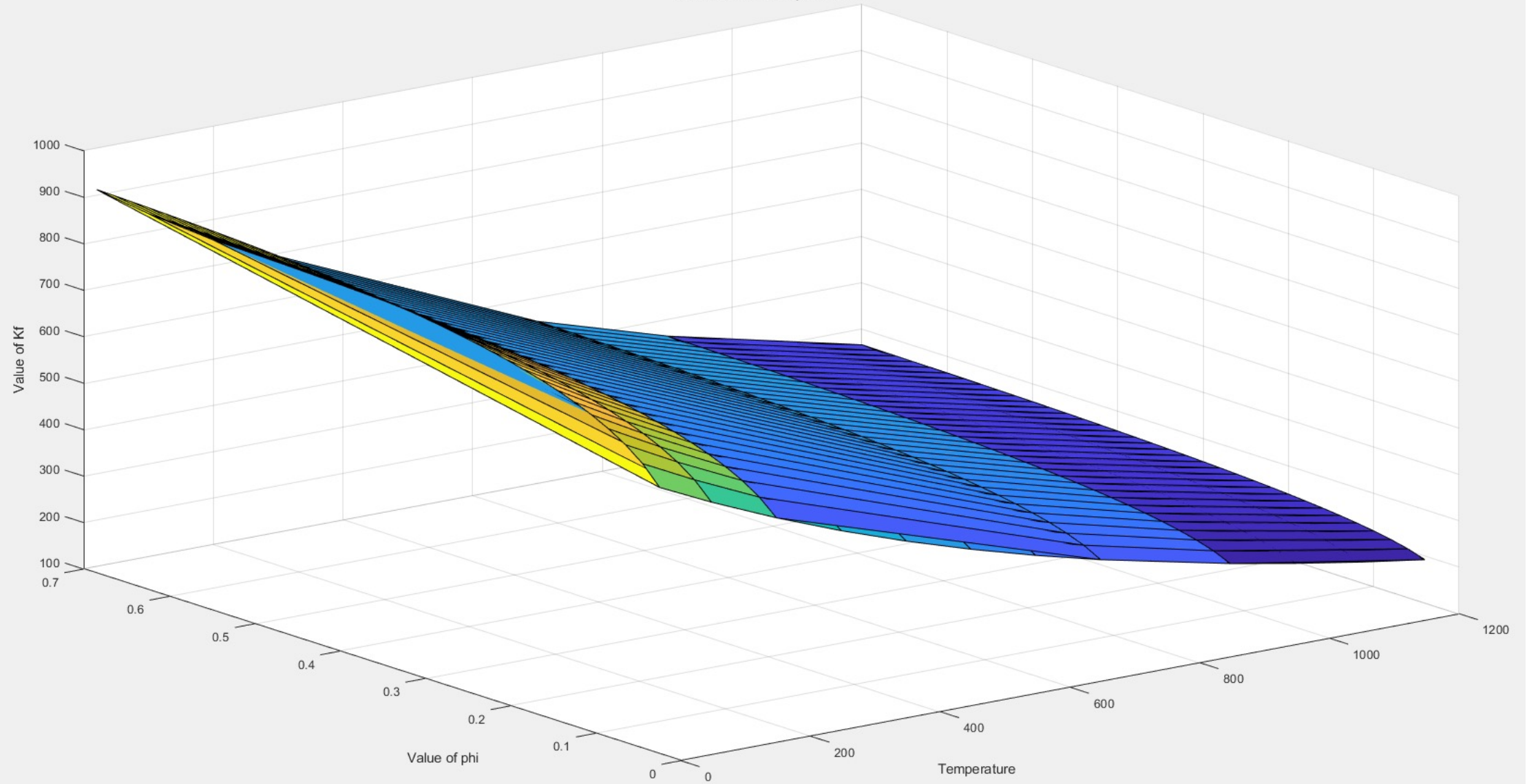




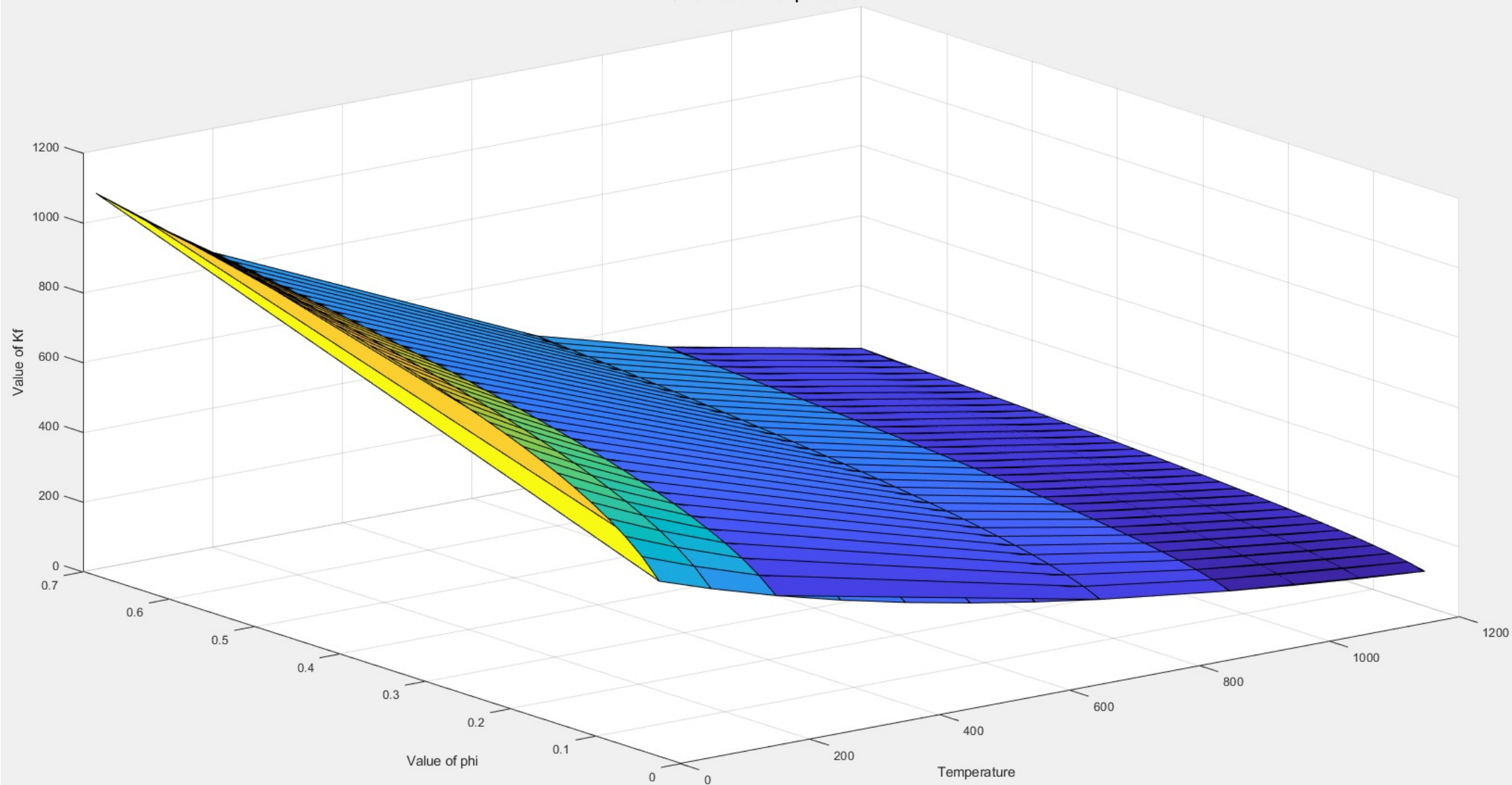
# 3 Dimensional graphs

For the 3 dimensional graphs we choose Phidot to be constant as it changes even less than Temp throughout the datasets

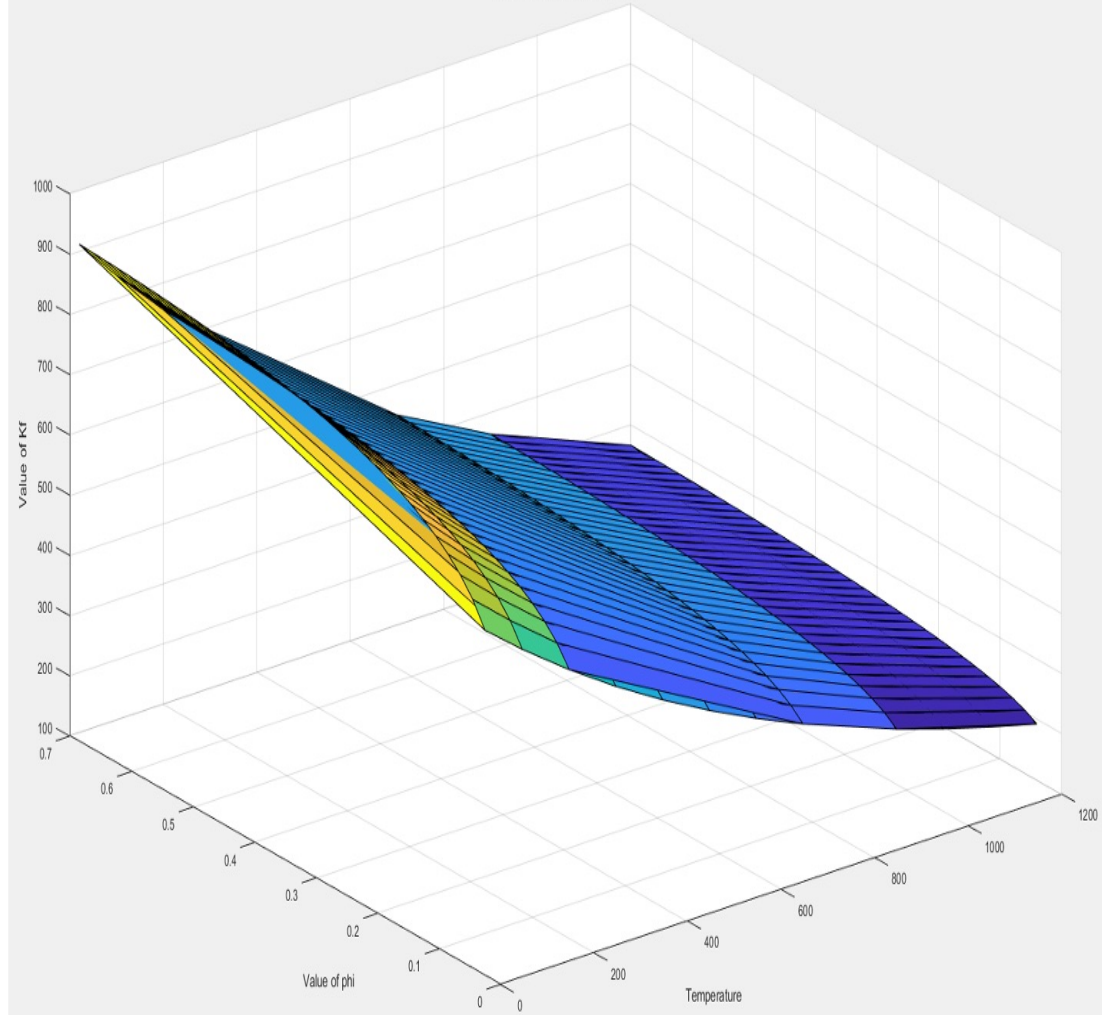
3D of model 1 with  $\phi_{dot} = 0.1$



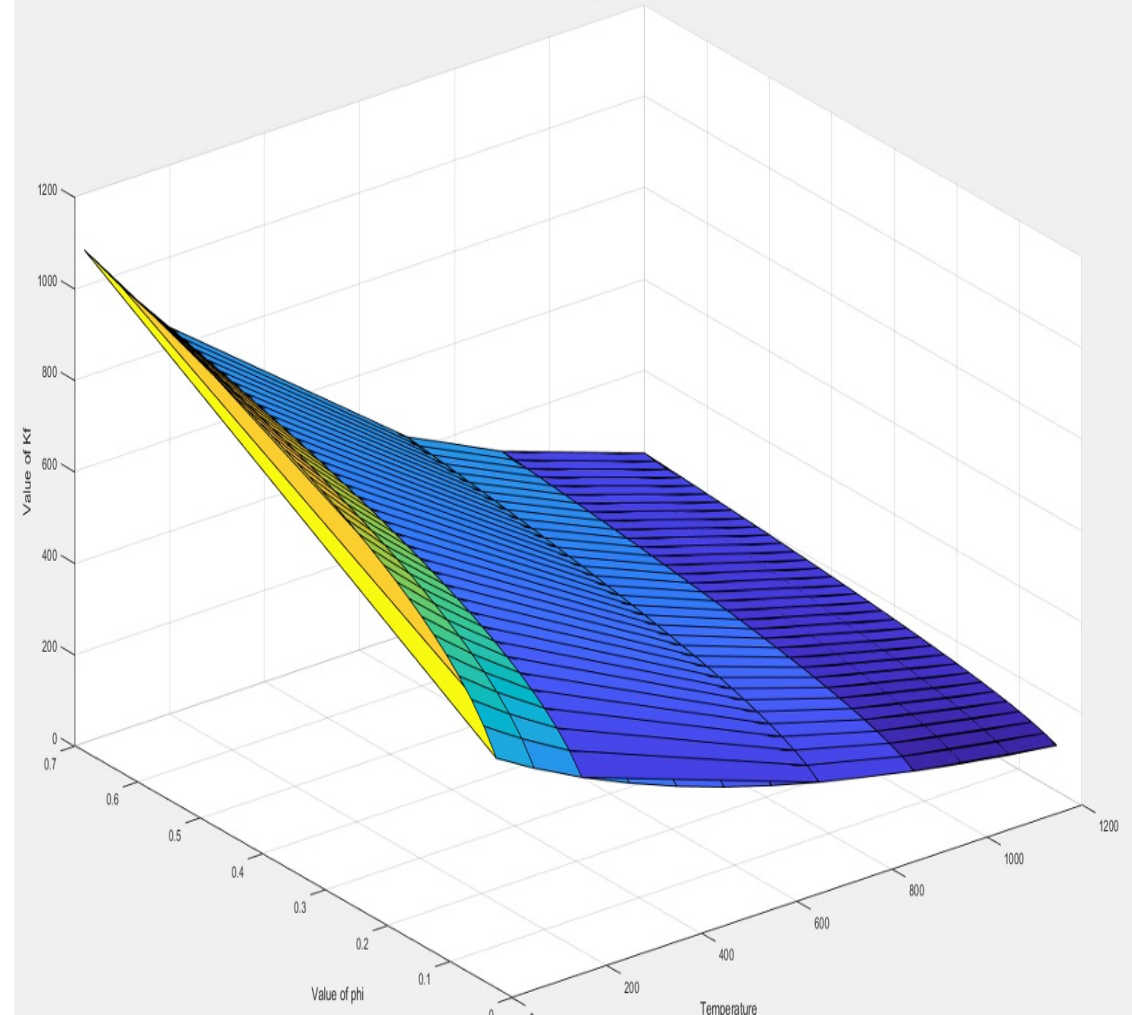
3D of model 2 with  $\phi_{dot} = 0.1$



3D of model 1 with  $\text{phidot} = 0.1$



3D of model 2 with  $\text{phidot} = 0.1$



# Comparison conclusion

We can conclude that values predicted by model 2 are closer to the kf values from the dataset at the lower values of phi but as phi increases model 1 becomes closer and closer to the kf value from the dataset



Thank you very much

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