

長晶製程參數智能化解析—— 斷線率改善與控制

智慧化企業整合

Intelligent Integration of Enterprise

- Project 3
- 108034541 周郁淇



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01

Scenario

- Background
 - 5W1H
- Literature Review

Background

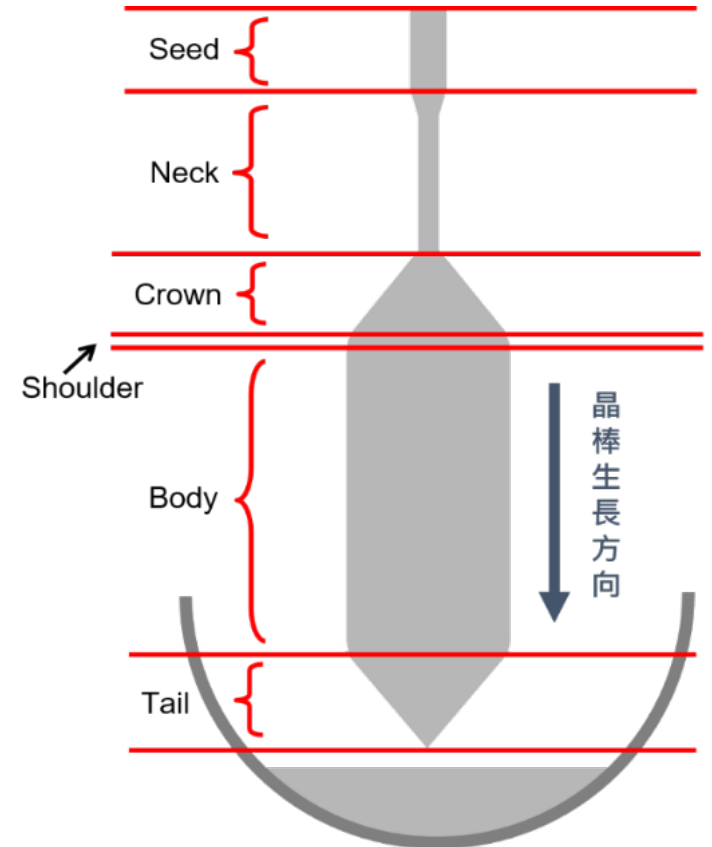
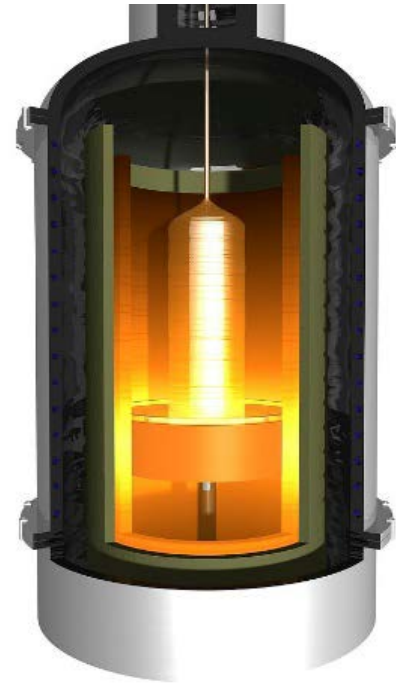
- 長晶製程

- 填料、熔融、拉起
- 晶頸→晶冠→晶身→晶尾

- 斷線

- 晶線：晶棒結晶的稜線
- 斷線：晶線不連續的現象

- 斷線重熔生長需1天以上工時
- 造成人力、設備成本消耗



5W1H

What

長晶製程斷線率
晶棒良率

W

When

晶棒生長的過程

W

Who

晶圓廠作業員
半導體公司

W

Where

晶圓廠長晶爐

W

Why

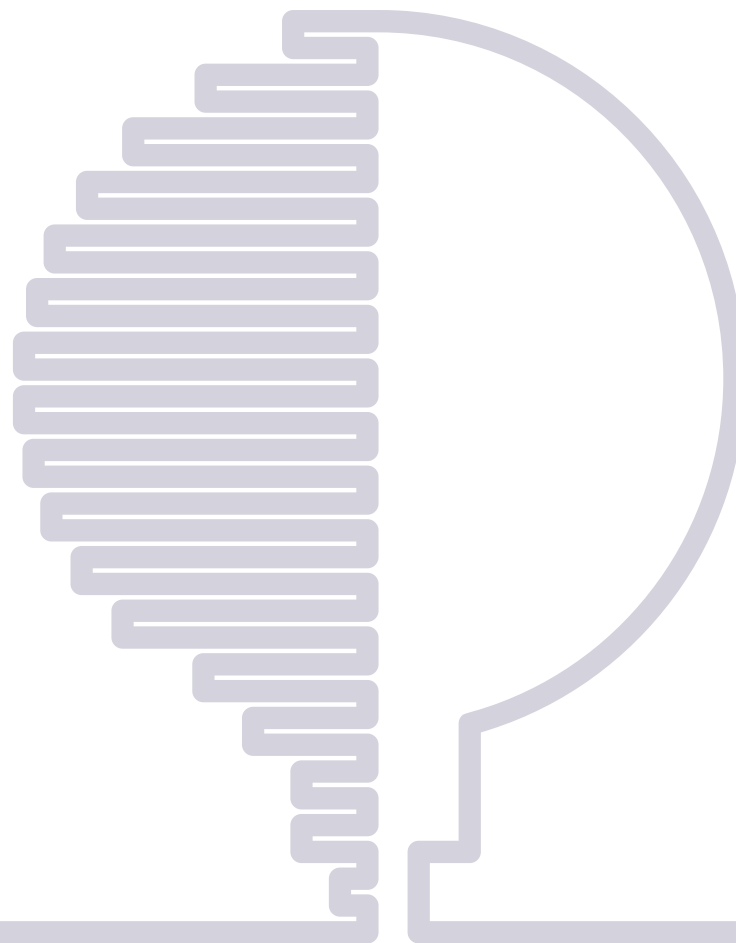
生長參數調整不當
晶棒斷線造成企業
成本增加

W

How

資料分析、深度學習、
機器學習

H



Literature Review

- 2018工業工程展專題
 - 長晶製程參數智能化解析——斷線率改善與控制
 - 迴歸分析參數數據(Minitab)
- 2019工業工程展專題
 - 半導體長晶製程斷線之檢測與改善
 - 迴歸分析參數數據(R)
 - 卷積類神經網路分析影像資料(CNN)
- 2019智慧化企業整合
 - 類神經網路分析參數數據(NN)



02

Machine Learning model

- Data-preprocessing
- Model architecture

Data-preprocessing



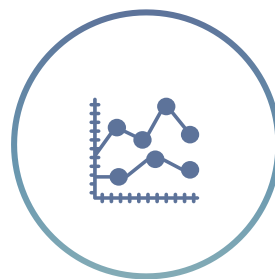
資料取得

晶圓廠非公開資料
長晶爐控制機台輸出



資料混和

98%斷線發生在晶身
非斷線：斷線 = 3:1
非斷線=0；斷線=1
取晶身長至55公分



關鍵少數

可控參數
目標拉速、目標晶棒直
徑、實際晶轉、實際坩
升比、實際氬氣流量、
上下爐腔爐壓



差分

計算參數變化量
 $f(t)=f(t)-f(t-1)$

Model architecture

- 類神經網路NN

```
from keras.models import Sequential
from keras.layers import Dense

classifier = Sequential()

classifier.add(Dense(input_dim = d, output_dim = 32 , init = 'uniform', activation = 'relu'))

classifier.add(Dense(output_dim = 16, init = 'uniform', activation = 'relu'))

classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'relu'))

classifier.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])

classifier.fit(train_set, y_train, batch_size=50, nb_epoch=50, validation_split = 0.1)
```

Accuracy
85%



03

Analysis

- Training process
- Training summary

Training process

- Optimizer

```
classifier = Sequential()

classifier.add(Dense(input_dim = d, output_dim = 32 , init = 'uniform', activation = 'relu'))

classifier.add(Dense(output_dim = 16, init = 'uniform', activation = 'relu'))

classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'relu'))

classifier.compile(optimizer = 'Nadam', loss = 'binary_crossentropy', metrics = ['accuracy'])

classifier.fit(train_set, y_train, batch_size=50, nb_epoch=50, validation_split = 0.1)
```

Accuracy
86%

- Activation function

```
classifier = Sequential()

classifier.add(Dense(input_dim = d, output_dim = 32 , init = 'uniform', activation = 'relu'))

classifier.add(Dense(output_dim = 16, init = 'uniform', activation = 'sigmoid'))

classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'sigmoid'))

classifier.compile(optimizer = 'Nadam', loss = 'binary_crossentropy', metrics = ['accuracy'])

classifier.fit(train_set, y_train, batch_size=50, nb_epoch=50, validation_split = 0.1)
```

Accuracy
90%

Training process

- Layers

```
classifier = Sequential()
classifier.add(Dense(input_dim = d, output_dim = 32 , init = 'uniform', activation = 'relu'))
classifier.add(Dense(output_dim = 16, init = 'uniform', activation = 'sigmoid'))
classifier.add(Dense(output_dim = 16, init = 'uniform', activation = 'sigmoid'))
classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'sigmoid'))
classifier.compile(optimizer = 'Nadam', loss = 'binary_crossentropy', metrics = ['accuracy'])
classifier.fit(train_set, y_train, batch_size=50, nb_epoch=50, validation_split = 0.1)
```

Accuracy
91%

- Neurons

```
classifier = Sequential()
classifier.add(Dense(input_dim = d, output_dim = 512, init = 'uniform', activation = 'relu'))
classifier.add(Dense(output_dim = 256, init = 'uniform', activation = 'sigmoid'))
classifier.add(Dense(output_dim = 256, init = 'uniform', activation = 'sigmoid'))
classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'sigmoid'))
classifier.compile(optimizer = 'Nadam', loss = 'binary_crossentropy', metrics = ['accuracy'])
classifier.fit(train_set, y_train, batch_size=50, nb_epoch=50, validation_split = 0.1)
```

Accuracy
93%

Training summary

Model	Optimizer	Activation function	Layers	Neurons	Accuracy
1	adam	relu	3	32,16,1	85
2	Nadam	sigmoid	3	32,16,1	86
3	Nadam	sigmoid	4	32,16,16,1	90
4	Nadam	sigmoid	4	512,256,256,1	93



04

Discussion

- Conclusion

Conclusion



改善斷線率

- 參數調整邏輯改善
 - 準確度93%
-

實務驗證

- 實驗設計
 - 即時監控
-

模型改善

- 資料擴充、處理
- 機器學習、深度學習

Thanks.

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