

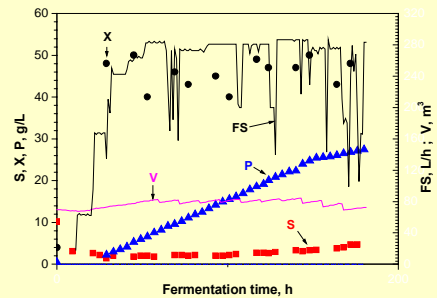
## State prediction and optimal scheduling for fed-batch fermentation

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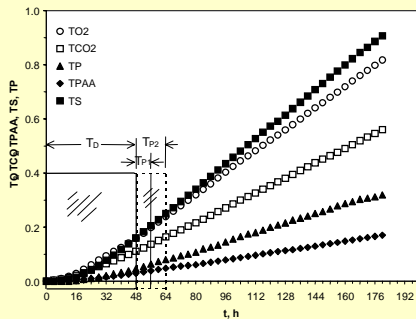
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- Development of software prediction technology
- Testing results
- Prediction of the profit function
- Online classification and optimal scheduling
- Demonstration of the software

## Typical time course of penicillin fed-batch fermentation



## Integrated substrate consumption & product formation



**Input:** discretized process variables within  $T_D$

**Output:** variables to be predicted up to 5-steps-ahead  
step length: 8h for antibiotic fermentation

**Generation of input-output data-pairs:** MDW-technique

**Training database:**  $\theta = \{\theta_{t-n}, \theta_{n+1}\}$

**Most important factors (Yuan and Vanrolleghem, JB, 1999):**

- selection of historical batches
- time span covered by historical batches
- off-line and online updating of the database
- detection and remediation of bad local minima

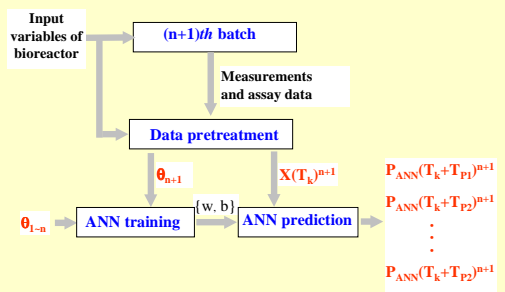
The  $k$ th data-pair  $[X(T_k), Y(T_k)]$  is presented by

$$X(T_k) = \begin{bmatrix} T_k \\ x(T_k) \\ x(T_k - 1\tau) \\ x(T_k - 2\tau) \\ \vdots \\ x(T_k - m\tau) \end{bmatrix}$$

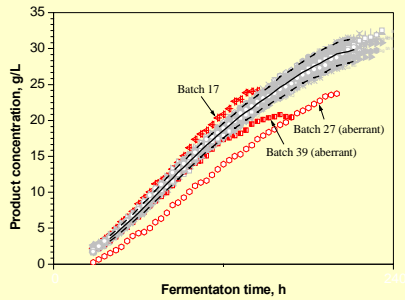
$$x(T_k) = [TO_2(T_k) \quad TCO_2(T_k) \quad TP(T_k) \quad TPAA(T_k) \quad TS(T_k) \quad \dots]^T$$

$$Y(T_k) = [P(T_k + T_{P1}) \quad P(T_k + T_{P2}), \dots, P(T_k + T_{P5})]^T$$

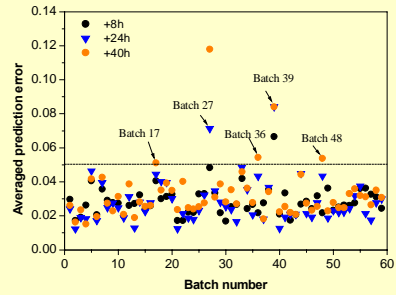
## Rolling learning-prediction procedure



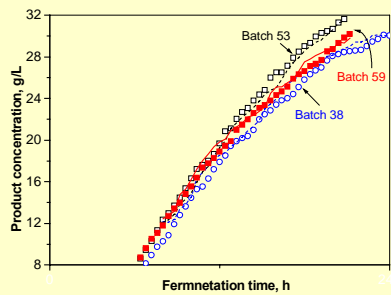
### Industrial fed-batch fermentations for testing of ANN-predictor



### Prediction accuracy (prediction of product concentration as an example)



### 40h-ahead prediction of product concentration

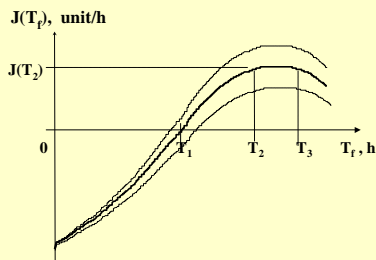


### Applications of the predicted state variables

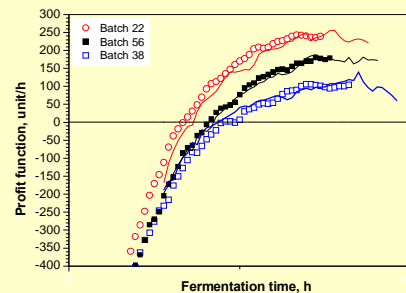
- Improvement of daily process supervision
- Early diagnosis of abnormal batches (indicated by unusual high prediction errors)
- Sub-optimal predictive control of feeding rates based on the predicted precursor/substrate uptake rates
- Estimation of economic potential for individual batches which may lead to optimal scheduling

### Profit function and optimal scheduling

**Profit function:**  $J(T_f) = \frac{\text{Revenue} - \text{Total costs}}{T_f + T_p}$

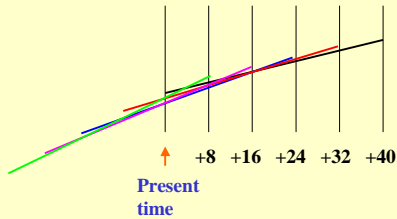


### 40h-ahead predicted profit function in comparison with its measurement

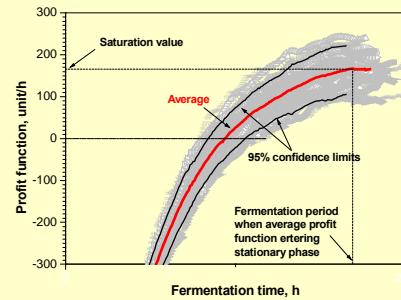


Scheduling is done during last quarter of  $T_f$

Using **average predicted value**  $J(t,t+40)_{AP}$  to indicate the future tendency:



Using **average measured value**  $J(t,t+40)_{AM}$  of the profit function as basis



Classification of batches according to economic potential

If  $\|J(t,t+40)_{AP} - J(t,t+40)_{AM}\| > J(t,t+40)_{AM} \times (1+\beta)$   
Then the **Good** category

If  $\|J(t,t+40)_{AP} - J(t,t+40)_{AM}\| < J(t,t+40)_{AM} \times (1-\beta)$   
Then the **Bad** category

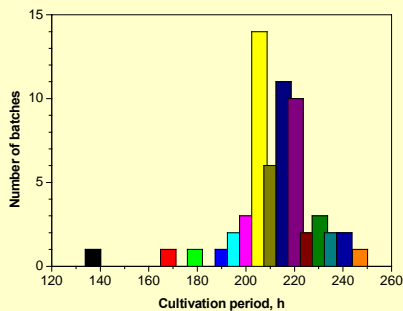
Otherwise, the **Normal** category

$\beta$  - significance factor, confidence limit- related  
 $\beta = 0.35 \sim 0.40$  for the testing data set

Scheduling rules for fed-batch cultivation

Category	Scheduling strategy
<b>Good</b>	Maintain operation as long as possible
<b>Normal</b>	Terminating as planned
<b>Bad</b>	Break off at an earlier time point

Feasibility of online optimal scheduling under "golden recipe"



Software demonstration given by Mr. Liu Jun