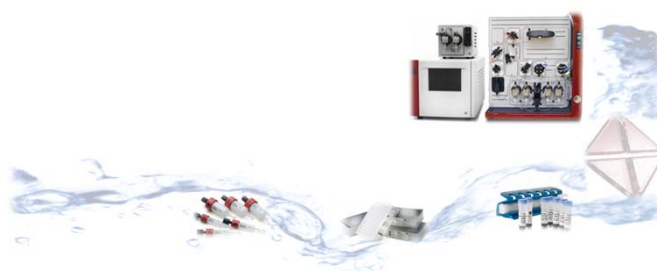


# Design of Experiments (DoE) in chromatography 层析中的实验设计(DOE)



## Content内容

**Introduction**介绍

**DoE Methodology** DoE方法学

**DoE in chromatography** 层析中的DoE

**Examples** 应用实例



# Introduction

## 介绍



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## What is Design of Experiments (DoE)? 什么是DoE?

**Processes depend on multiple factors!**  
工艺基于多个因子（变量）

Statistical  
multifactorial  
approach  
统计学的多因子  
实验方法

Maximize  
information  
最大的信息量

Minimize  
experiments  
最少的实验数量

Detect  
interacting  
factors  
发现交互作用的  
因子



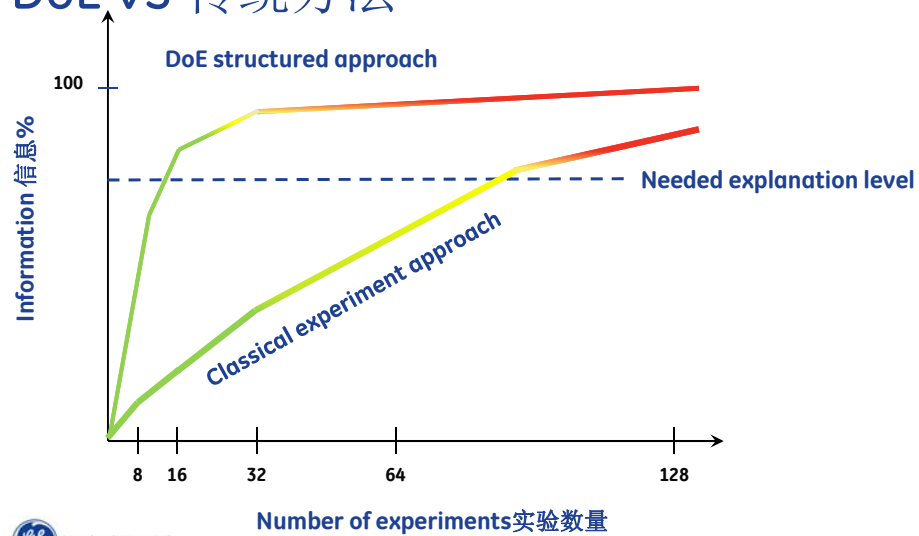
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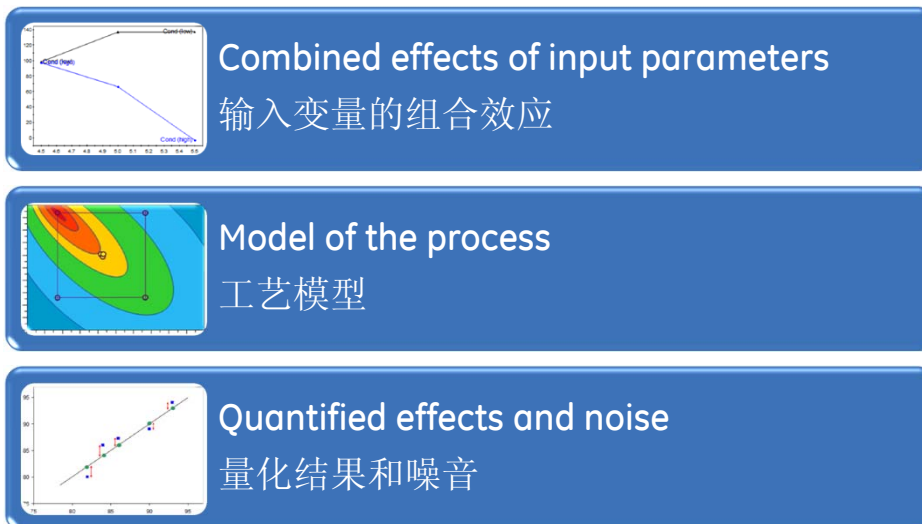
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## DoE vs. Classical approach DoE VS 传统方法



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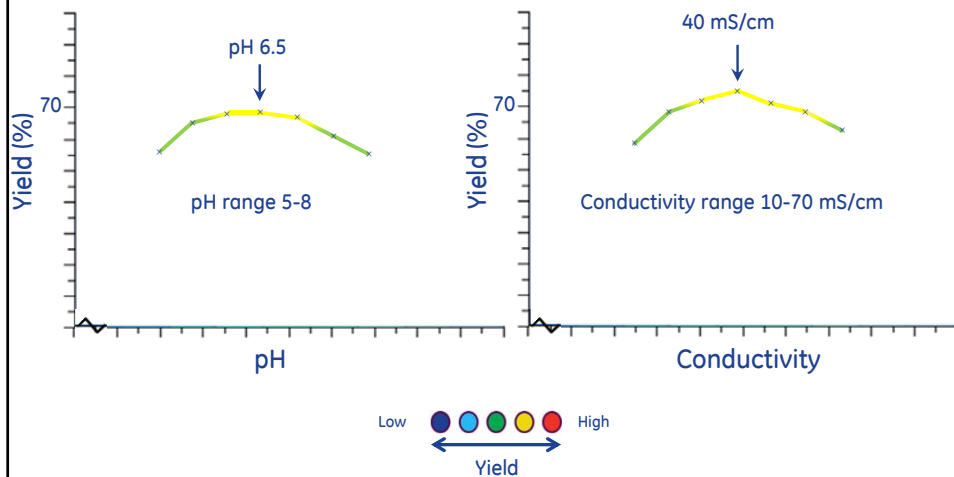
## Output 输出



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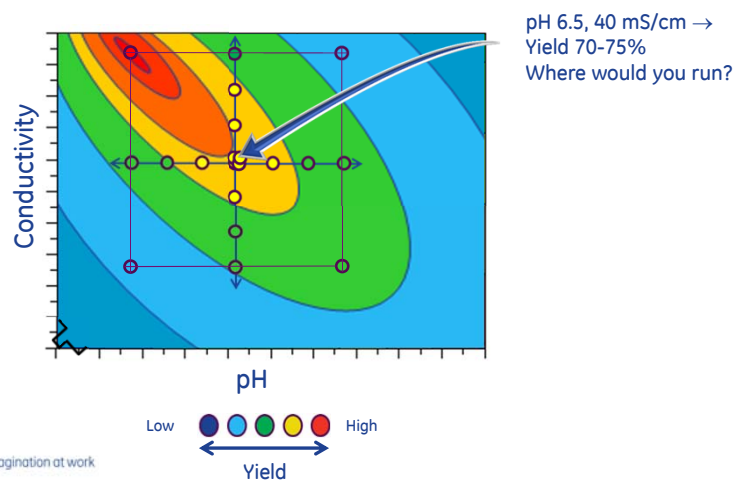
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## One Factor At a Time 单因子实验



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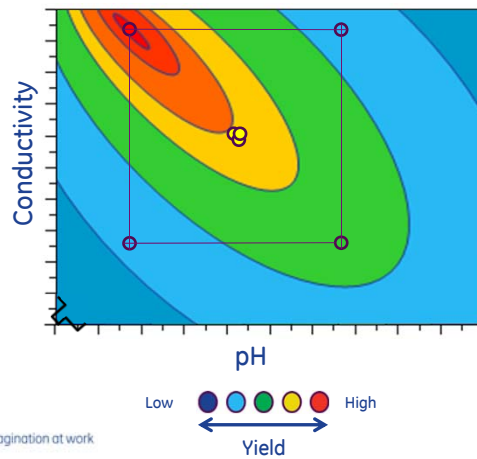
## One Factor At a Time vs. DoE 单因子实验 VS DoE



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## One Factor At a Time vs. DoE 单因子实验 VS DoE

Yield of the reaction is influenced by pH and conductivity! pH和温度共同影响收率



A two-factor interaction effect  
两因子的交互作用

- The influence of one factor depends on the setting of another factor  
一个因子的效应取决于另一个因子
- Two-factor interactions are common in chemistry  
两因子交互作用在化学中很常见

One could imagine three-factor interactions but they are rarely (never) significant.

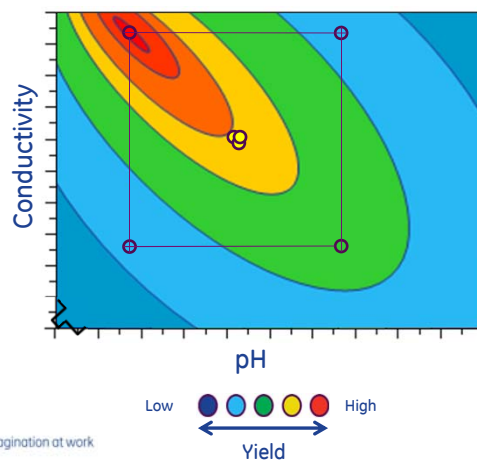
当然也会有三因子交互作用，但是几乎不显著甚至不存在

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## One Factor At a Time vs. DoE 单因子实验 VS DoE

Detection and quantification of interactions i.e. a true optimum can be found !

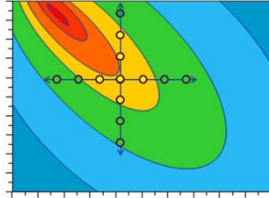
发现并且量化交互作用 才能找到真正的最优条件



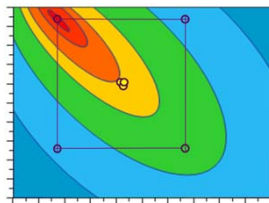
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## Advantages of DoE over OFAT

### DoE对OFAT的优势:



VS.



OFAT → assumes response is a linear combination of the critical factors!

OFAT: 假定响应是关键因子的线性组合

OFAT → doesn't explore the whole experimental region!

OFAT: 没有探索整个实验区间

OFAT → risk for sub optimization, and/or unpleasant surprises!

OFAT: 存在局部优化和不愉快的意外风险!

DoE → fewer resources for the amount of information obtained

DoE: 更少的资源获得大量的信息

DoE → estimates of the effects of each factor are more precise

DoE: 更精准地评估各个因子的影响

DoE → systematic estimation of interactions

DoE: 系统化地评估交互作用

DoE → experimental information from entire factor space

DoE: 整个因子空间的实验信息

DoE → obtain transfer functions to identify the optimum solution

DoE: 获得传递函数, 可以确定最优的结果



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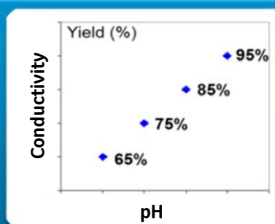
Box, Hunter, and Hunter (1978) *Statistics for Experimenters*, New York: Wiley.  
Czitrom, Veronica, *The American Statistician*, May 1999, Vol. 53, No. 2 pp 126-131

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## Aliasing – something to avoid 混淆: 需要避免的情况

Example: Does Temperature, pH or both affect Yield?

例如: 温度, pH 还是两者共同影响收率?



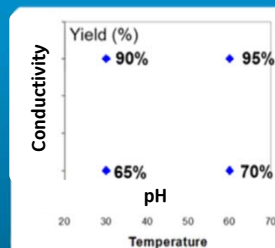
### Alternative 1 结果1

pH and Conductivity are said to be **ALIASED** (confounded, correlated)

如果pH和温度被认为是混淆的(困惑的, 关联的)

It is impossible to say if the effect we see on yield is due to pH, Conductivity or both!

无法确定对收率的影响是来自于温度, pH还是两者



### Alternative 2 结果2

pH and Conductivity are **NOT ALIASED** (non-confounded, uncorrelated)

如果pH和温度是不混淆的(不困惑的, 不关联的)

We clearly see that both factors affect Yield, and that the effect from Conductivity is largest!

可以清楚地发现两个因子都影响收率, 但是pH的影响更加显著

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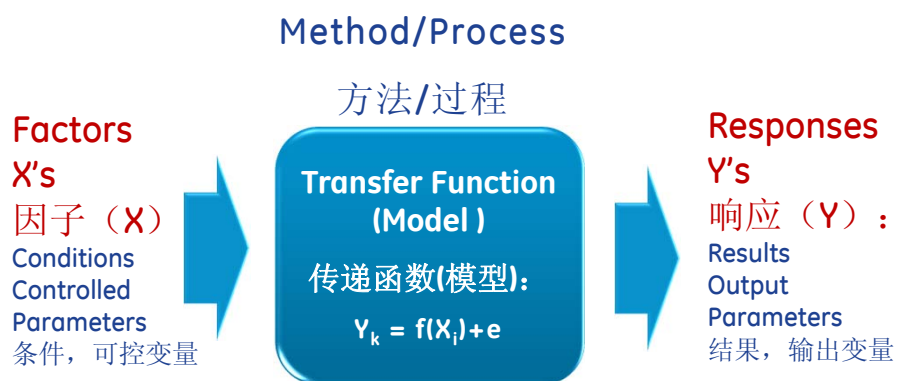
# DoE Methodology

## DoE方法学



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### DoE terminology DoE的术语:



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## Transfer Function (Model)

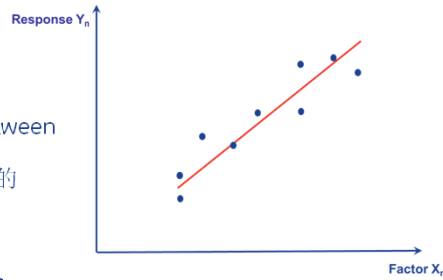
型) 传递函数 (模  
 $y = f(x) + e$

$y$  Measured responses  
 可检测的响应

$x$  Factor (parameter) settings  
 设定的因子 (参数)

$f(x)$  The transfer function/model  
 quantifies the relationship between factors and responses  
 定量描述因子和效应之间关系的传递函数 (模型)

$e$  Residuals=  
 Random experimental variation  
 Quantifiable in a DoE  
 残差=可在DoE中量化的随机的实验变量



## Transfer Function (Model)

传递函数 (模型)

Response Y<sub>n</sub>

Minimized errors between the measured data and the teorethical data calculated according to the model

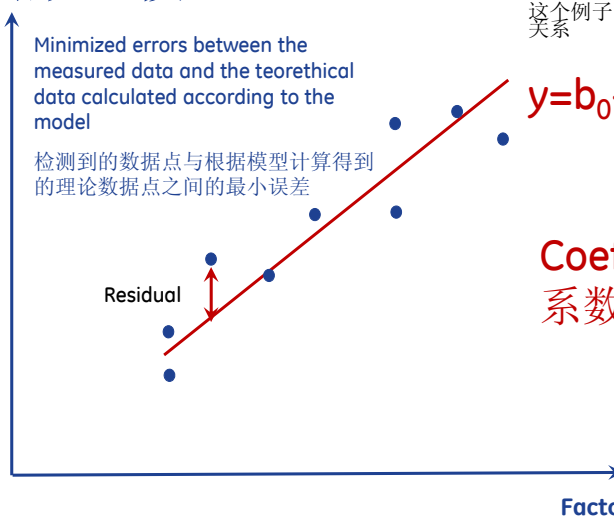
检测到的数据点与根据模型计算得到的理论数据点之间的最小误差

Residual

The "model", in this case is a linear relationship  
 这个例子中的模型时线性关系

$$y = b_0 + b_1 x_1 + e$$

Coefficients  
 系数





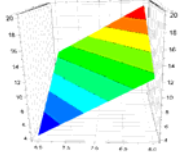
## Transfer Function (Model) 传递函数(模型)

Linear Terms  
(main effects)  
线性项 (主效应)

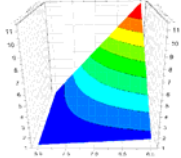
Interaction Term(s)  
交互项

Quadratic Term(s)  
二次项

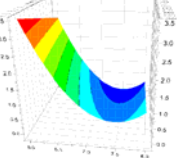
$$y = b_0 + b_1x_1 + b_2x_2 + b_{12}x_1x_2 + b_{11}x_1^2 + b_{22}x_2^2 + e$$



Robustness/Screening  
稳健性/筛选



Screening/Optimization  
筛选/优化



Optimization  
优化

$b_1 - b_2 - b_{12} - b_{11} - b_{12}$  coefficients give the quantified effects for the  $x$ 's 这些为传递函数的系数，是参数 $x$ 被量化的影响。

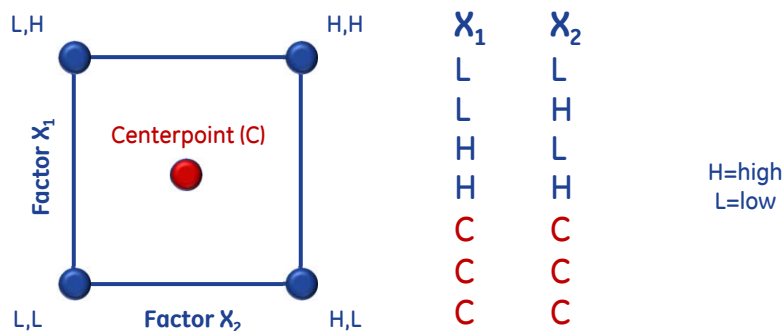
$e$  = prediction error  
预测误差  
 $y_{\text{predicted}} - y_{\text{measured}}$



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## General design construction 一般的设计思路



Centerpoint used for estimation of noise and detection of curvature  
中心点用于评估误差，发现曲率

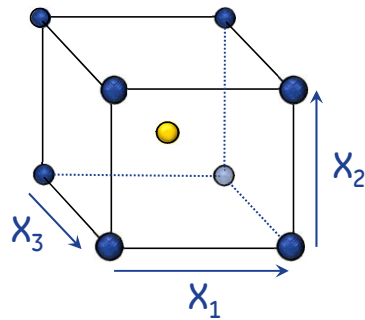


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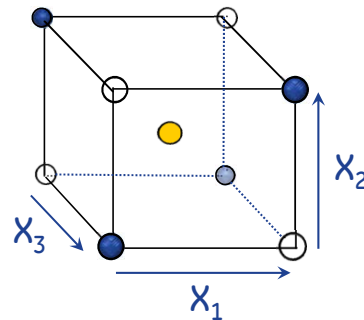
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## General design construction

### 一般的设计思路



Full Factorial design  
完全析因设计



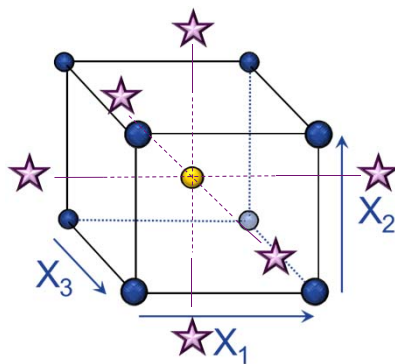
Fractional Factorial design  
部分析因设计



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## Central Composite Designs (CCD's)

### 中心复合设计



#### Corner Points拐点

The factorial part of the design.  
For the assessment of linear and 2-way interaction terms.

设计的因子部分，评估线性和2次交互的常数项

#### Center Points中心点

Used to detect curvature. 用于检测曲率

Replicated to estimate pure error  
重复点用于估计纯误差

#### Star Points星点

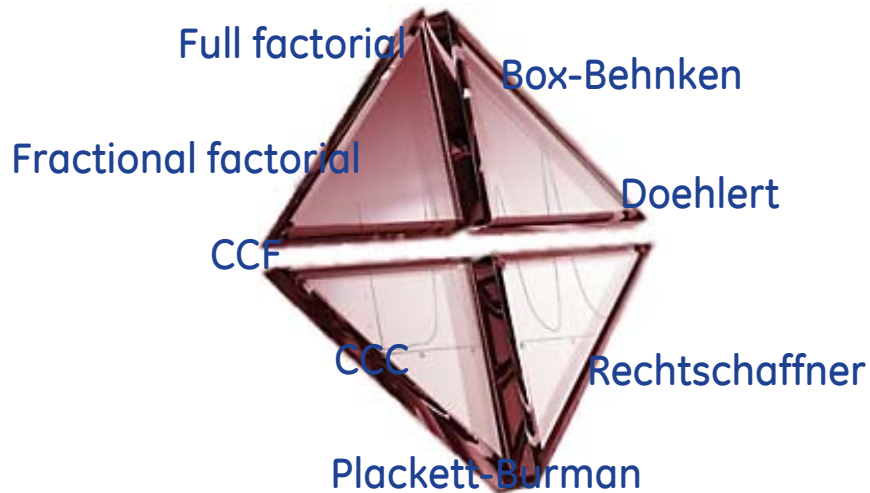
For the assessment of quadratic terms, quantify curvature.

用于评估二次项的常数项，定量曲率

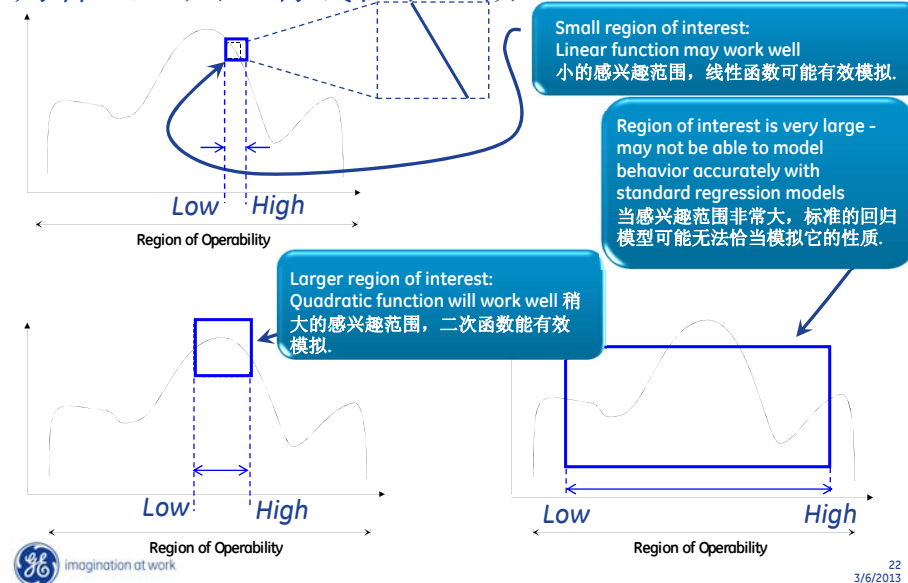


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## Which design? 选择何种设计方法

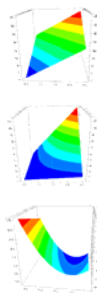


## Why (only) linear and quadratic fits? 为什么（只）有线性和二阶拟合？



## How many runs in a design?

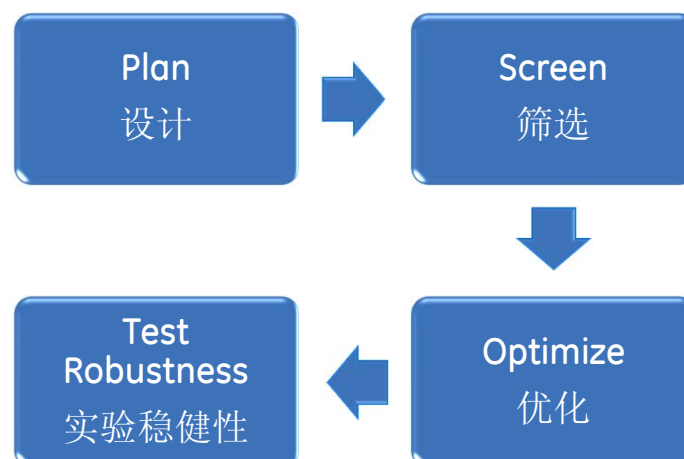
一次设计需要多少次实验？



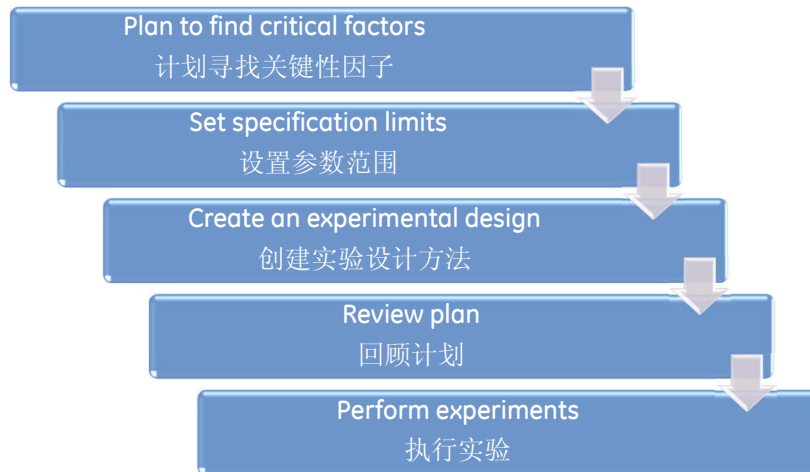
# of runs for some common experimental designs	2	3	4	5	6	7
Fractional factorial res III	--	7	--	11	11	11
Fractional factorial res IV	--	--	11	--	19	19
Fractional factorial res V	--	--	--	19	35	65
Rechtschaffner res V	--	10	14	19	25	32
Full factorial	7	11	19	35	67	131
Central composite RSM	11	17	27	29	47	81
Rechtschaffner RSM	--	13	18	24	31	39
Box-Behnken RSM	--	15	27	43	51	59
Doehlert RSM	9	15	23	33	45	59

- Res III = limited support for linear effects, no support for 2-way interactions and quadratics
- 分辨度为III的设计= 有限支持线性效应（各主效应之间没有混杂），不支持二阶交互效应和二次项
- Res IV = good support for linear effects, limited support for 2-way interactions, no quadratics
- 分辨度为IV的设计= 能很好地支持线性效应，有限地支持二阶交互效应，不支持二次项
- Res V = good support for linear and 2-way interaction effects, no quadratics
- 分辨度为V的设计= 能很好地支持线性效应和二阶交互效应，不支持二次项
- Full factorial = good support for linear and all order of interactions, no quadratics
- 全因子= 能很好地支持线性效应和全部交互效应，不支持二次项
- RSM = Optimization designs, good support for linear, 2-way interaction and quadratic effects
- 响应面= 优化设计，能很好地支持线性效应，二阶交互效应和二次项

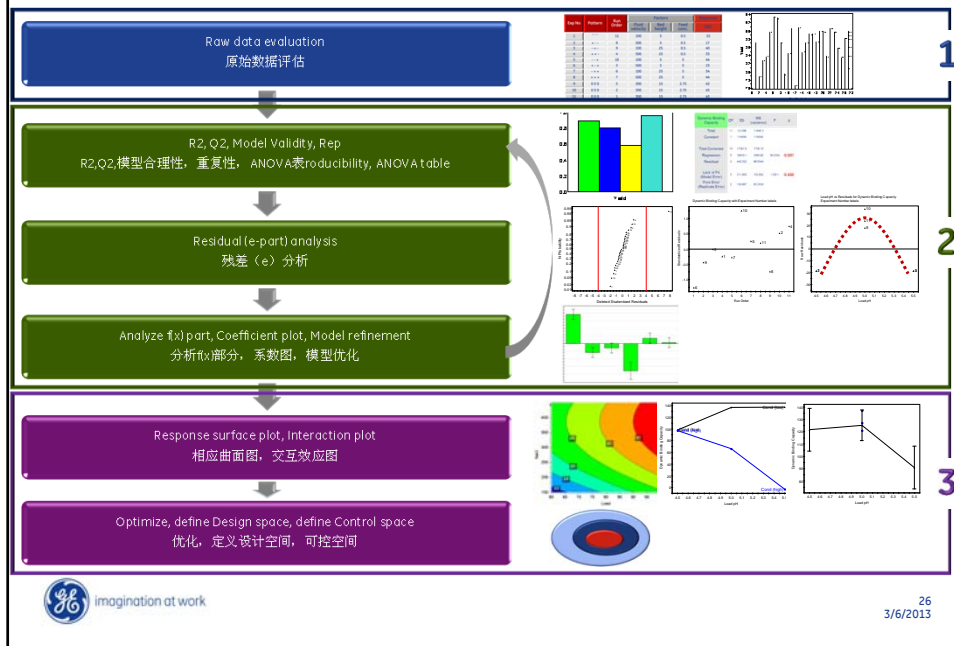
## DoE approach DoE的方法



## In practice....实践中

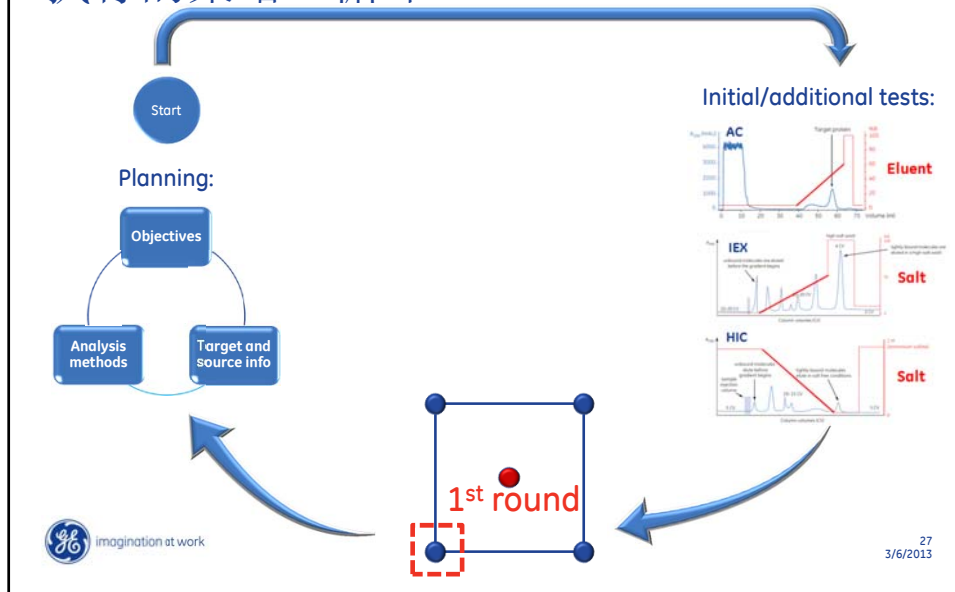


## DoE evaluation workflow DoE评估流程



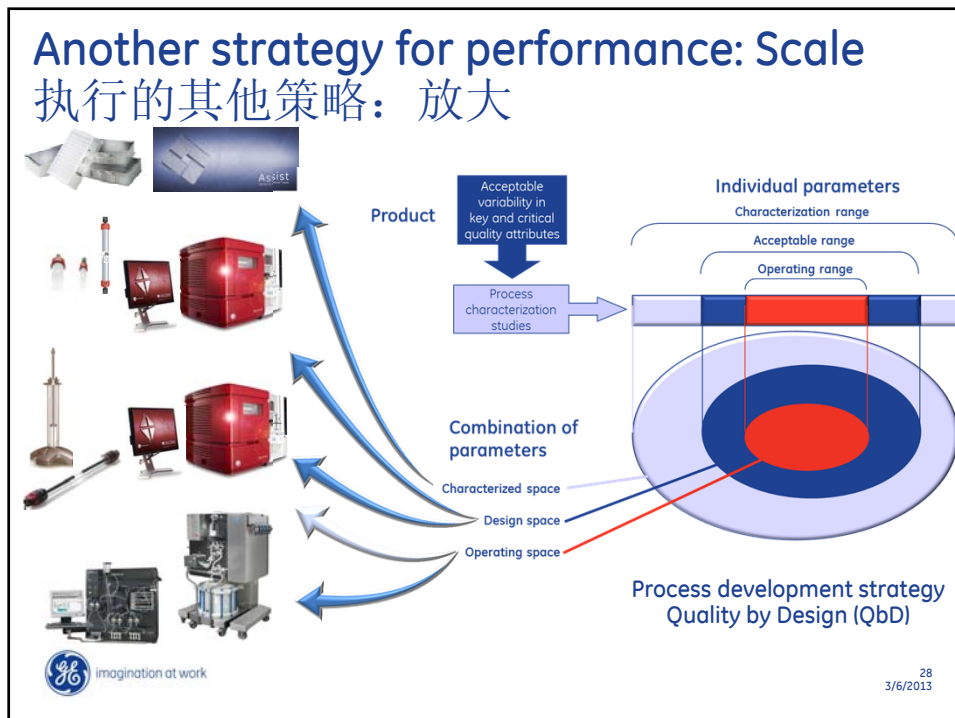
## Strategy for performance: Iteration

### 执行的策略：循环



## Another strategy for performance: Scale

### 执行的其他策略：放大



## DoE in chromatography 层析中的DoE



### Purification protocol development 纯化方案的开发



How many purification steps is required 需要多少纯化步骤呢?

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# Protein purification and analysis

## 蛋白纯化和分析

### Protocol development

#### 开发方案

- Confirm correct protein

确认正确表达蛋白

- Target protein concentration

目标蛋白浓度

- Activity活性

- Protein solubility蛋白稳定性

- Confirm correct protein

确认正确表达蛋白

- Target protein concentration

目标蛋白浓度

- Activity活性

- Purity纯度

- Size homogeneity大小均一性

- Protein stability蛋白稳定性

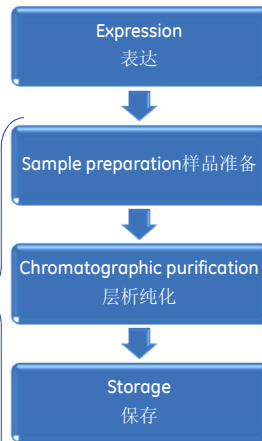
- Protein modifications蛋白修饰

- Application specific analysis

特定应用的分析方法



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### Routine process

#### 常规步骤

- Target protein concentration

目标蛋白浓度

- Target protein concentration

目标蛋白浓度

- Purity纯度

- Size homogeneity 大小均一性

- Confirm correct protein确认正确表达蛋白

- Activity活性

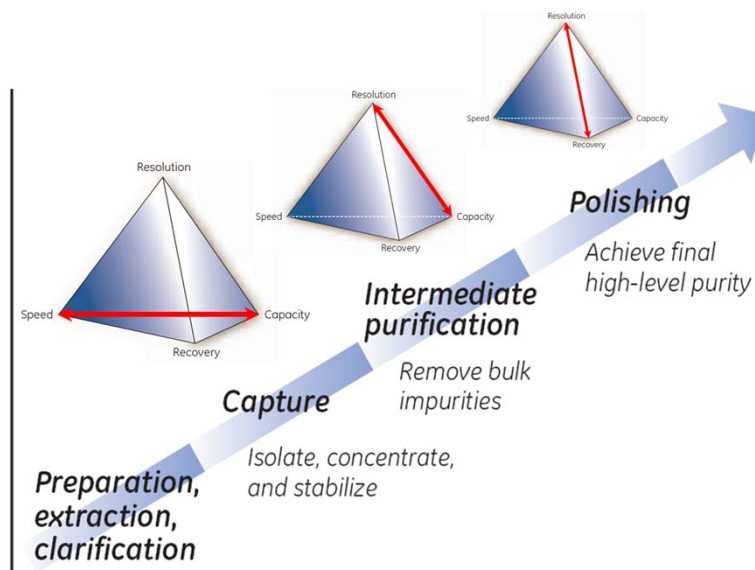
- Application specific analysis

特定应用的分析方法

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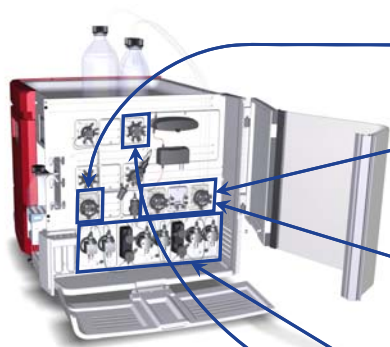
## Apply CiPP Strategy!应用CiPP策略!



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## Factors and responses in chromatography 层析中的因子和响应：



Sample conditions  
上样条件

Wash conditions  
清洗条件

Elution conditions  
洗脱条件

Entire process  
整个过程

### Factors

Load pH  
Load conductivity  
Load concentration  
Mass load  
Wash volume  
Wash pH  
Wash conductivity  
Elution pH  
Gradient elution  
Step elution level  
Cut OD  
Elution Additive  
Media type  
Column size  
Bed Height  
Flow rate  
Residence time

### Responses

#### External data:

Capacity  
DBC (Frontal analysis)  
Yield  
Purity/Selectivity  
Molecular weight  
Activity  
HCP  
DNA  
Aggregates  
Protein A

#### Peak Data:

Area  
Concentration  
Amount  
Resolution  
Asymmetry  
Plates per meter



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## Performing DoE in chromatography 层析中使用DoE

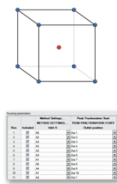
### UNICORN method

Method Settings  
Equilibration  
Sample Application  
Column Wash  
Elution  
Equilibration

### Design input

- Definition of factors, factor types and settings
- Definition of objective for creating the design

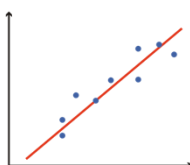
### Design and scouting



### Run



### Model evaluation



### Use of model for prediction and decisions



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## Example: optimization study 实例：优化研究



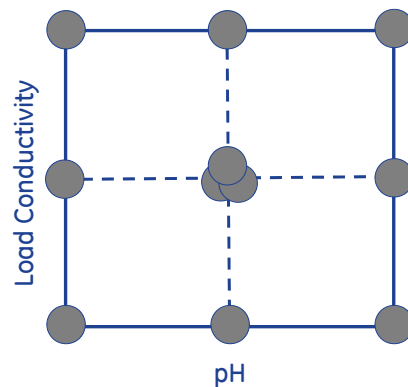
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## Optimization of loading conditions 上样条件的优化

- Optimization of MAb sample loading conditions on Capto S

优化Capto S上MAb样品的上样条件

Exp No	Exp Name	Run Order	Load pH	Load Conductivity
1	N1	4	4.5	5
2	N2	10	4.5	15
3	N3	3	5.5	5
4	N4	11	5.5	15
5	N5	7	5	5
6	N6	1	5	15
7	N7	5	4.5	10
8	N8	9	5.5	10
9	N9	2	5	10
10	N10	6	5	10
11	N11	8	5	10



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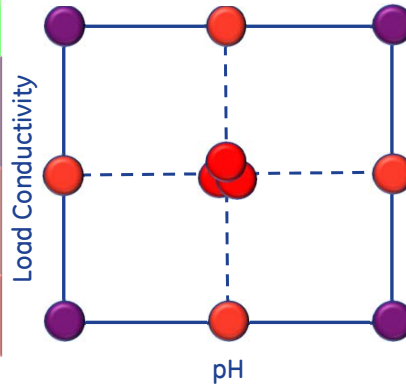
## Evaluate CCF design CCF设计的评估:

Model:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$

Our design supports quadratic terms

我们的设计支持二次项

Exp No	Exp Name	Run Order	Load pH	Load Conductivity	Dynamic Binding Capacity
Factorial part			4.5	5	96
			4.5	15	102
			5.5	5	137
			5.5	15	4
Star points			5	5	139
			5	15	54
			4.5	10	119
			5.5	10	84
Replicated Center Points			5	10	121
			5	10	137
			5	10	127

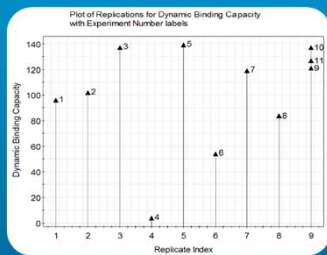


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## Replicate plot of the raw data

原始数据的重复点图



Replicate variation vs. variation in entire experimental design.

重复变量 VS 整个实验设计中的变量

Replicate variation ok? 重复变量OK?

Response target range? 响应的目标范围?

Suspicious values? 可疑点?

2<sup>nd</sup> degree curvature? 二次曲率?



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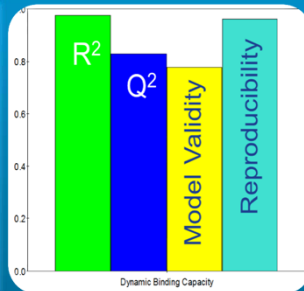
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## Summary of Fit 拟合的总结

Model 模型:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$



Model describes the observed variation in DBC

模型描述了 DBC 实验中所有可检测到的变量

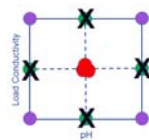
Excellent R<sup>2</sup> & very good Q<sup>2</sup>. 极好的 R<sup>2</sup> & Q<sup>2</sup>

Model validity: residual variation < reproducibility of center points, i.e. no Lack of Fit.

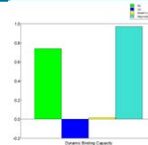
模型的有效性: 残差的变化 < 中心点的重复性

Reproducibility bar: Center point variation low compared to overall variation

重复性: 比较总变异和中心重复点间变异计算得到的



Bad model →  
差的模型



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## ANOVA table ANOVA 表

Model:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$

Dynamic Binding Capacity	DF	SS	MS (variance)	F	p
Total	11	131398	11945.3		
Constant	1	114036	114036		
Total Corrected	10	17361.6	1736.16		
Regression	5	16919.1	3383.82	38.2334	0.001
Residual	5	442.522	88.5044		
Lack of Fit (Model Error)	3	311.855	103.952	1.5911	0.408
Pure Error (Replicate Error)	2	130.667	65.3334		

ANOVA statistical tests ANOVA 统计学检验

Current model OK! 现有模型 OK!

Variance in model (Regression) > Variance not captured (Residual) → R<sup>2</sup> value is significant

模型 (回归方程) 方差 > 剩余方差 → R<sup>2</sup> 显著

Model Error (residuals excluding replicate variance) is NOT significantly greater than the Replicate Error i.e. the model has No Lack of Fit. 模型误差 (除掉重复试验的剩余方差)

不明显大于重复误差, 模型不失拟。

Dynamic Binding Capacity	DF	SS	MS (variance)	F	p
Total	7	87944	12563.4		
Constant	1	74882.3	74882.3		
Total Corrected	6	13061.7	2176.95		
Regression	3	9674.75	3224.92	2.85646	0.206
Residual	3	3386.96	1128.99		
Lack of Fit (Model Error)	1	3256.3	3256.3	49.8413	0.019
Pure Error (Replicate Error)	2	130.667	65.3334		

Bad model  
Significant lack of fit  
差的模型  
丢失重要的拟合项



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## Conclusions so far 目前的结论

Model:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$

Statistically significant portion of the observed variation in DBC explained.

DBC中检测变量的统计学显著性得到解释

Variation in model residuals (the model error) is NOT significantly larger than the variation in the replicated center points

模型残差方差（模型误差）不显著大于中心重复点方差。

Our current model is of sufficient complexity to describe the process.

目前的模型有足够的复杂度来描述该工艺。

→ **Next step: Evaluate the residual variation.**

下一步，评估剩余方差



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## Analyzing the residuals 残差分析

Residuals = observed - predicted for each experiment 残差=观测值-实验预测值

Ideally, the model error (residuals) will consist of normally distributed random variation from the experimental process and measurement systems. IF the residuals are NOT normally distributed, one should try to identify the reason why. 理想情况下，实验过程和分析系统中模型错误（残差）会服从正态分布的随机变化。如果残差不是正态分布，必须分析确定原因

**Tool工具:**

**Normal probability plot of the residuals** 残差正态分布图

Also, there should be no trends visible when looking at the model error versus either the run order in which the experiments were performed or versus any of the X variables. 另外，在残差 vs 实验顺序或 vs 任何X变量时，应该无明显的趋势性

**Tools工具:**

**Residuals vs run order plot** 残差 vs 实验顺序

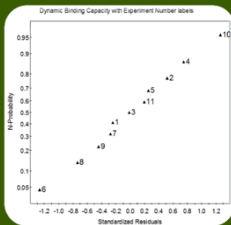
**Residuals vs individual X variables** 残差 vs X变量



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# Normal probability plot of residuals 残差正态分布图

Model:  $DBC = k + b_1 * pH + b_2 * Cond + b_{11} * pH^2 + b_{22} * Cond^2 + b_{12} * pH * Cond + e$



Forms a nice straight line with no outliers or groupings in the data  
形成一条漂亮的直线，并且没有异常值或者组群的值。

Implies residuals are normally distributed. 表明残差是正态分布

Describing the experimental noise 描述了实验噪音。

All experiments used to provide information on the random variability of a process.

所有的实验用于为工艺中的随机变量提供信息

Model uncertainty  $\sim \pm 2 * RSD$ . Less near the center of the investigated experimental space, more towards the outer limits of the X parameters. 模型的不确定性  $\sim \pm 2 * RSD$ . 与考察的实验区间中心越远，越靠近X的外部极限

Non-normally distributed (random)

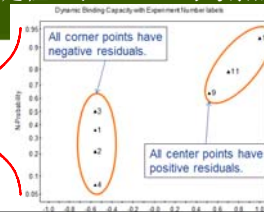
非正态分布（随机）

Systematic variation

系统偏差



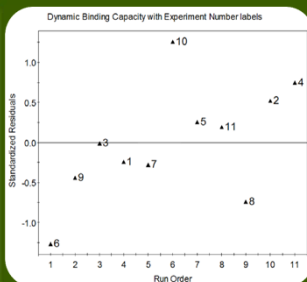
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# Residuals vs run order 残差 vs 实验顺序

Model:  $DBC = k + b_1 * pH + b_2 * Cond + b_{11} * pH^2 + b_{22} * Cond^2 + b_{12} * pH * Cond + e$



There are no strong trends in the residuals when plotted vs the order in which the experiments were performed 按照实验顺序作图时，没发现残差有明显

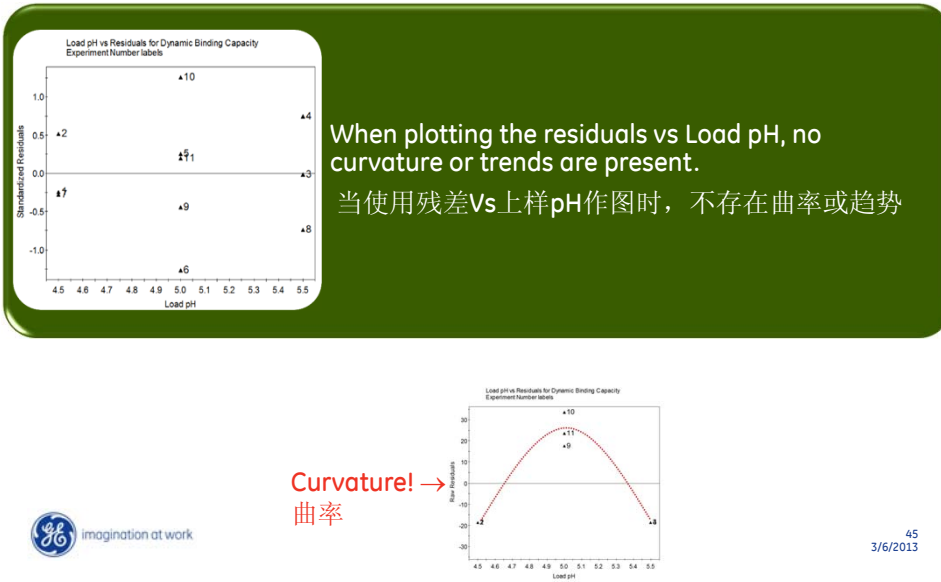


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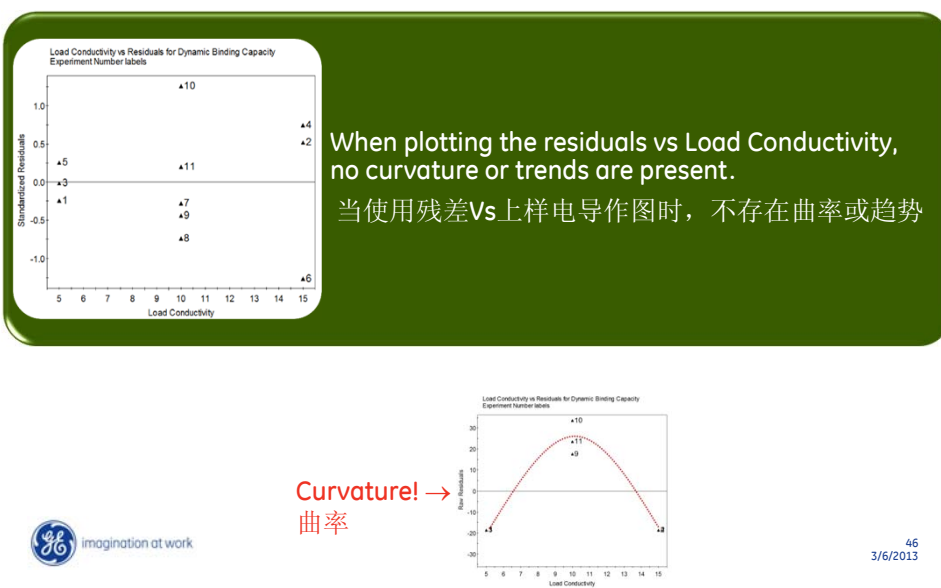
## Residuals vs Load pH 残差 vs Load pH

Model:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$



## Residuals vs Conductivity 残差 vs 电导

Model:  $DBC = k + b_1 \cdot pH + b_2 \cdot Cond + b_{11} \cdot pH^2 + b_{22} \cdot Cond^2 + b_{12} \cdot pH \cdot Cond + e$



## Conclusions so far 目前的结论

Model:  $DBC = k + b_1 * pH + b_2 * Cond + b_{11} * pH^2 + b_{22} * Cond^2 + b_{12} * pH * Cond + e$

The current model describes a statistically significant portion of the observed variation in Dynamic Binding Capacity (DBC)

目前模型具有统计学意义地描述动态载量中观测到的变化。

The variation in model residuals (the model error) is NOT significantly larger than the variation in the replicated center points.

模型方差中的变化没有比在中心重复点上的变化显著大很多

The residuals are normally distributed with no outliers or groupings in the data, therefore the residual standard deviation (RSD) can be used as a measure of the random variability of the process (including measurement system variation).

残差符合正态分布并且没有异常值或组群，因此剩余标准偏差可以用于估量在过程中的随机变化性（包括测量系统引起的变化）

→ We are now done with the "e" part of  $Y = f(X) + e$

我们现在分析完传递函数的误差部分

→ Next step: Evaluate the model, i.e. the  $f(X)$  part of  $Y = f(X) + e$

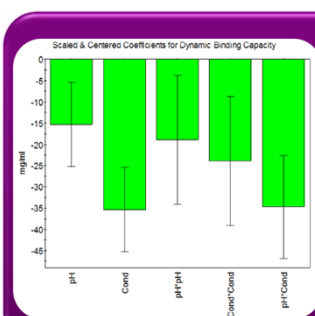
下个阶段：评估模型



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## Model coefficient plot 模型系数图

Model:  $DBC = k + b_1 * pH + b_2 * Cond + b_{11} * pH^2 + b_{22} * Cond^2 + b_{12} * pH * Cond + e$



Verify that all model terms are statistically significant.

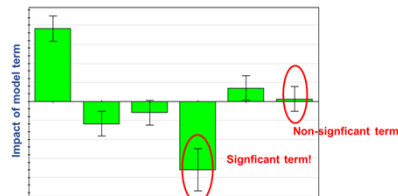
Statistically significant model terms (95% confidence level)

验证了所有的模型系数都是统计学显著(95%置信)

Model refinement: Remove non-significant model terms!  
模型修饰：去除不显著的模型条件



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## Interpreting the model 解释模型

Now that we have successfully evaluated and verified that the model is sound from a statistical perspective, we can look at interpreting the quantified cause and effect relationships for our investigated process.

现在我们已经成功从统计学观点评估并验证了模型的可行性，接下来可以解读量化的因果关系

Remember that statistical significance, while important, always should be considered from a domain expertise perspective. In other words, the DoE model should make sense.

记住，统计学意义虽然重要，但是总应该从一个专业领域的视角考虑。换句话说，DoE模型应该能被解释。

For the purpose of interpreting the model, the following tools are frequently used in DoE: 为了解模型，下面的工具常在DoE中用到：

**Interaction plots** 交互作用图

**Main effect plots** 主效应图

**Response surface plots** 响应曲面图

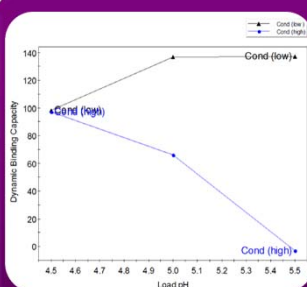


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## Interaction plots 1(2) 交互作用图

How the effect from pH on DBC depends on conductivity  
pH对DBC的影响取决于电导



Effect from Load pH on DBC depends on the Load conductivity.

上样pH对DBC的影响依赖于上样时的电导

Low conductivity (5 mS/cm) → increasing the pH results in an increased DBC, but at high conductivity cm) the effect from increasing the pH is a drastic DBC.

低电导(5ms/cm) → 增加pH会增加DBC；高电导时增加pH会使DBC急剧下降。

Note: this plot should only be interpreted for significant interaction effects.

注意：这条曲线仅仅解释有重要交互关系的影响因素

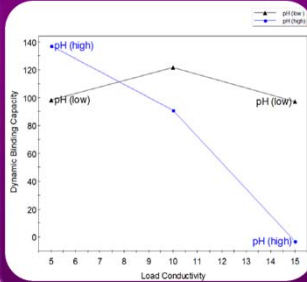


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## Interaction plots 2(2)交互作用图

How the effect from conductivity on DBC depends on pH  
电导对DBC的影响取决于pH



This plot shows how the effect from Load conductivity on DBC depends on the Load pH.

这张图显示了上样电导对DBC的影响依赖于上样pH

At low pH (4.5), increasing the conductivity results in a slightly increased DBC followed by a slight drop, but at high pH (5.5) the effect from increasing the conductivity is a drastic drop in DBC.

低pH时（4.5）：增加电导会使DBC先小幅增加，再小幅降低；高pH时，增加电导会使DBC急剧下降。

Note: this plot should only be interpreted for significant interaction effects.

注意：这条曲线仅仅解释有重要交互关系的影响因素

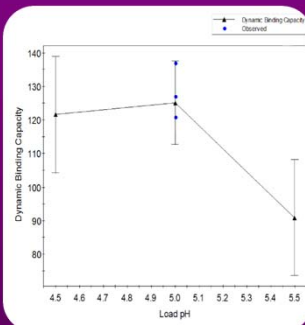


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## Main effect plots 1(2)主效应图

How pH affects DBC around the center point  
在中心点pH影响DBC



This plot shows the effect from Load pH on DBC at center point setting for Load conductivity.

这张图显示了在上样电导中心点时上样pH对于DBC的影响

The error bars indicate the width of the model prediction confidence intervals.

误差棒显示了模型预测值的置信区间的宽度。

The replicated center points are shown in blue.

重复点用蓝色显示

Note: this plot should mainly be used for X parameters involved in no or relatively small interaction effects, or in direct combination with the interaction plot(s).

注意：这个图主要用于该X变量没有或较小的参与到交互作用中



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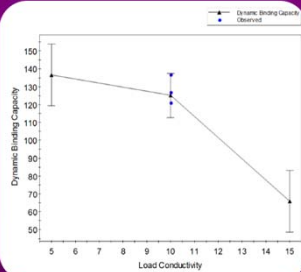


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## Main effect plots 2(2)主效应图

How conductivity affects DBC around the center point



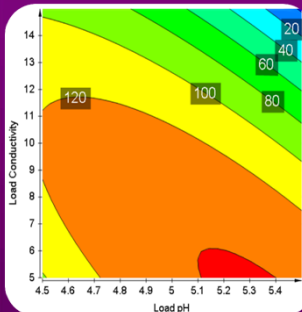
Similar to the previous slide, this plot shows the effect from Load conductivity on DBC at center point setting for Load pH.

与上张slide相似，这个图显示的是在中心点电导影响DBC。

## 2D Response surface plot

### 2D 相应曲面图

How DBC depends on both Load pH and Load Conductivity  
pH和电导对DBC的影响



This response surface shows the quantified effects, i.e. the  $f(X)$  part, from Load pH and Load Conductivity on DBC.

显示的是pH和电导对DBC定量的效应

Great tool for visualizing interaction and curvature effects. 可视化交互效应和曲率的工具

The residual standard deviation (RSD) should always be considered when looking at these plots.

当观察这些图时，剩余的标准偏差（RSD）要一直铭记于心

For this model,  $RSD = 9.4$ , i.e. the model uncertainty is roughly  $\pm 18.8$  mg/ml.

从这个模型看出， $RSD=9.4$ ，模型的不确定性大概是 $\pm 18.8$  mg/ml.

## Example: Robustness testing

### 实例：稳健性实验



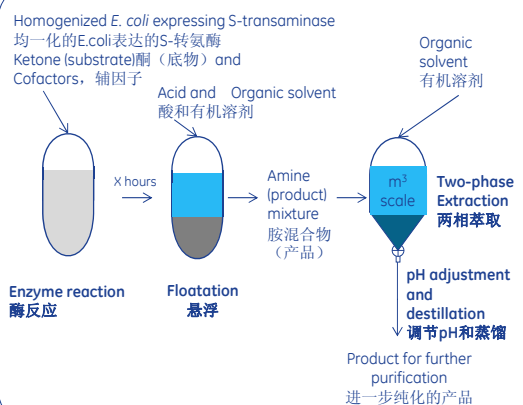
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## Robust NHS coupling?

### NHS偶联的稳定性实验？

#### Present process

##### 现有工艺



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#### New process

##### 新工艺



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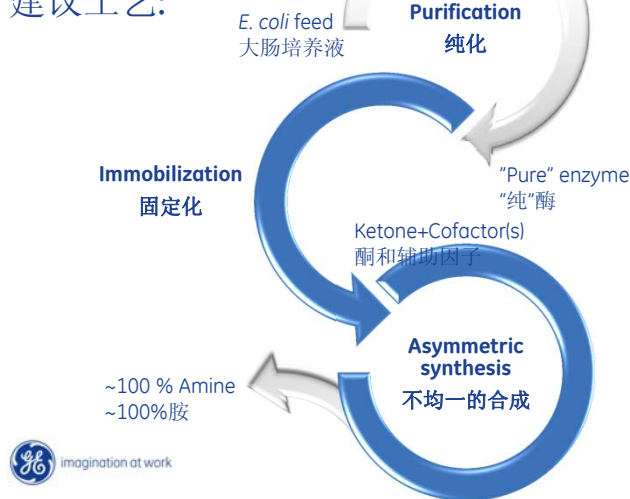


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## Robust NHS coupling?

## NHS偶联的稳定性实验?

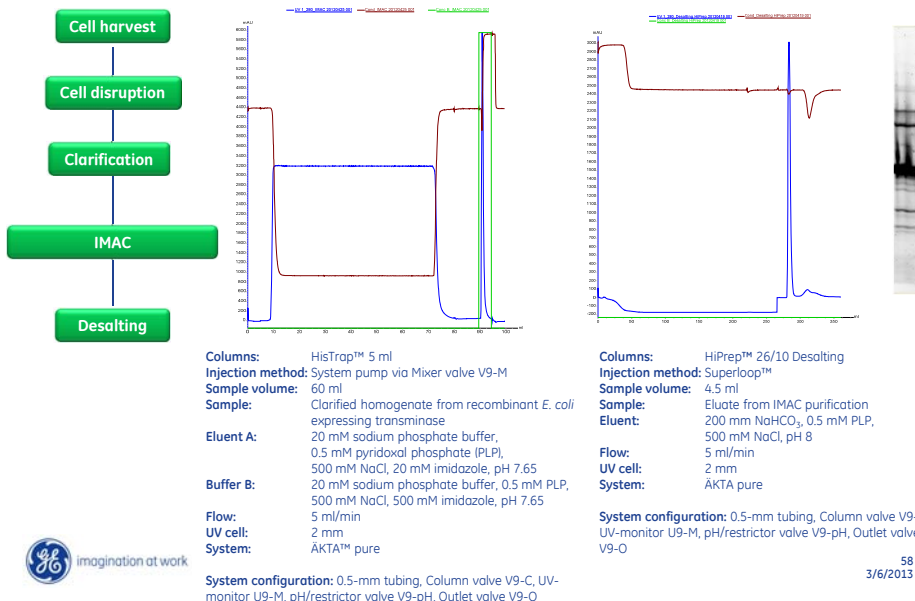
Proposed process  
建议工艺:



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## IMAC purification and desalting

## IMAC纯化和脱盐



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## Robust NHS coupling? NHS偶联稳定性试验? (Full Factorial完全析因)

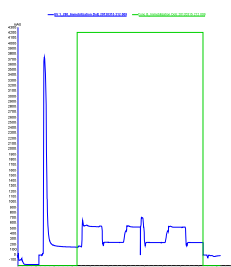
### Factors因子

Load 上样量: 3-5 mg  
Time 时间: 30-120 min  
pH: 8-9

### Response 响应

Coupling efficiency 偶联效率

Exp no	Run No	pH	Incubation time (min)	Load (mg/ml)
1	9	8	30	3
2	11	8	120	3
3	2	8	30	5
4	1	8	120	5
5	3	9	30	3
6	7	9	120	3
7	8	9	30	5
8	10	9	120	5
9	6	8.5	75	4
10	4	8.5	75	4
11	5	8.5	75	4

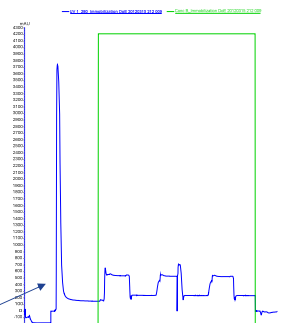


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## Scouting of NHS coupling conditions 优化NHS偶联的条件

Columns: HiTrap™ NHS-activated HP, 1 ml  
Injection method: Loop valve V9-L with 1-ml loops  
Sample volume: up to 0.7 ml  
Sample: Purified transaminase in carbonate buffer, pH 8.0, pH 8.5, pH 9.0  
Eluent A: 100 mM sodium phosphate buffer, 0.5 mM PLP, pH 7.5 (coupling buffer)  
Buffer B: 100 mM Tris-HCl, 0.5 mM PLP, pH 8.5  
Buffer C: 100 mM Acetate buffer with 0.5 mM PLP, 0.5 M NaCl, pH 4  
Buffer C: 1 mM HCl  
Flow: 1 ml/min  
UV cell: 2 mm  
System: ÄKTA™ pure

System configuration: 0.5-mm tubing, Column valve V9-C  
UV-monitor U9-M, Loop valve V9-L



N-hydroxysuccinimide group, which is released during the coupling reaction, shows strong absorbance at 280 nm (pH above 6). Clearly this peak can also be caused by protein that was not coupled. See column instructions for description of evaluation of binding efficiency. 偶联过程中释放的N-琥珀酰亚胺基团，在280nm有强吸收值(pH>6)，可以清晰地看到这个峰是由于没有偶联的蛋白造成的。利用这个峰来计算偶联率



### Method outline



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## Robustness check of NHS-coupling conditions

### NHS偶联的条件的稳定性试验检验

Scouting runs using ÄKTA™ pure

使用ÄKTA™ pure进行Scouting

#### Factors

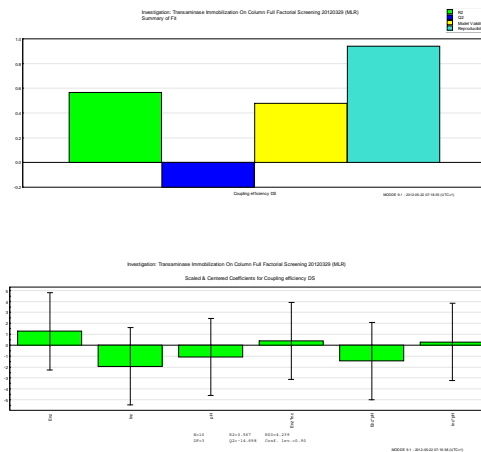
Load: 3-5 mg

Time: 30-120 min

pH: 8-9

#### Response

Coupling efficiency: 87.5-98.3 %



## Summary总结

Organized approach which connects experiments in a rational manner

有组织的方法，使得实验理性科学地进行

More information by investigating the influence of all factors together

通过同时考察所有因子的影响来获得更多的信息

More precise information is acquired in fewer experiments

更少的实验得到的是更加准确的信息

Results are evaluated in the light of variability

根据可变性来评估结果

Support for decision-making

决策强有力的支持



# Thank you!



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