

# The Use of Quality-by-Design and DOE Tools for BioAssay Development: Part 1: Component Optimization

Dr. Laureen E. Little

# Contact Information

**Laureen Little, Ph.D.**

Quality Services  
Principal Consultant

phone: 951-659-1957

email: [Biotech@ix.netcom.com](mailto:Biotech@ix.netcom.com)

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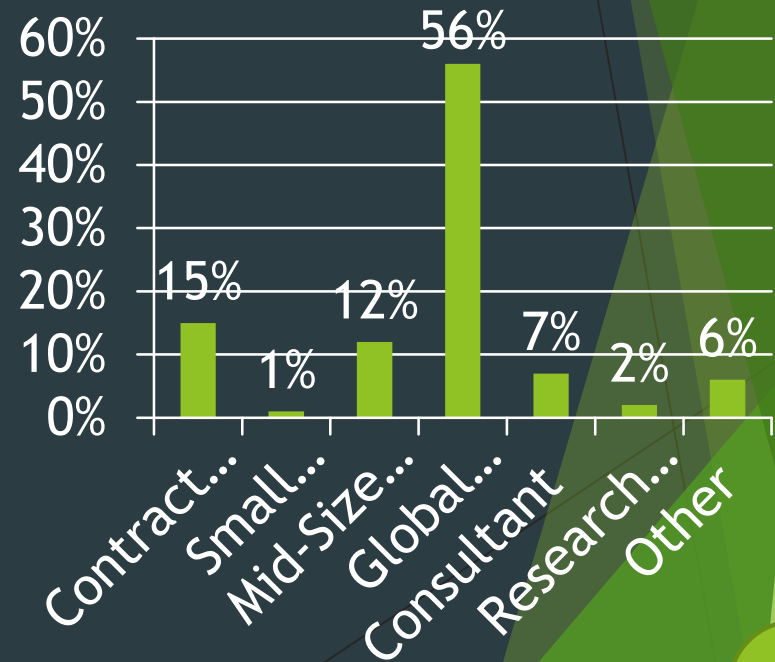
# Before we start:

- ▶ You have a transceiver. These are to allow us to do some interactive things.
- ▶ When the clock appears in the bottom right hand side push a number for your answer.
- ▶ A green light will appear. If it remains green and then goes out your answer was accepted. If the light becomes red your answer was not received. Try again.
- ▶ If you hit the wrong answer - just answer again. The first answer will be removed and replaced with the most recent answer. (only 1 answer allowed per transceiver.)

# About You

What Kind of Company do you work for?

1. Contract Organization
2. Small Biopharmaceutical (< 50 employees)
3. Mid-Size Biopharmaceutical (50 - 300)
4. Global Pharmaceutical
5. Consultant
6. Research Institute
7. Other

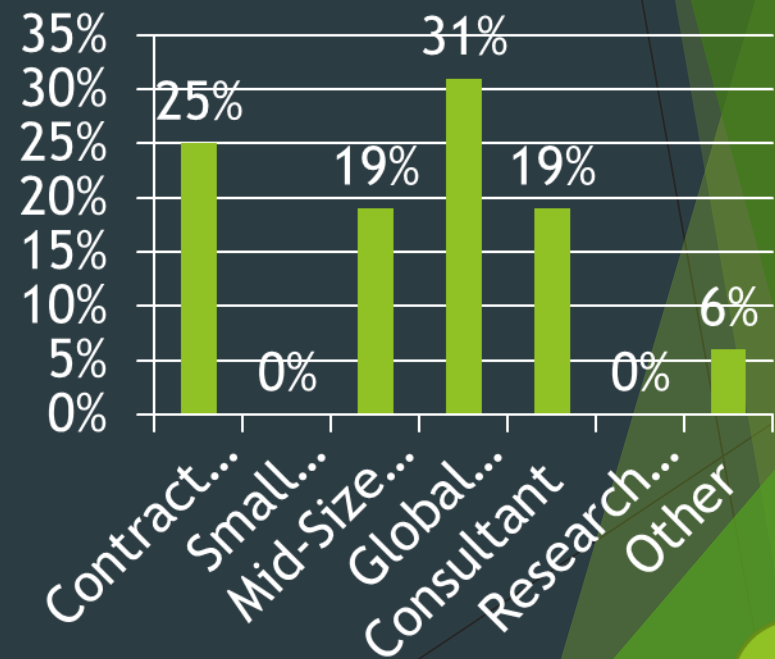


# About You

## EU RESULTS

What Kind of Company do you work for?

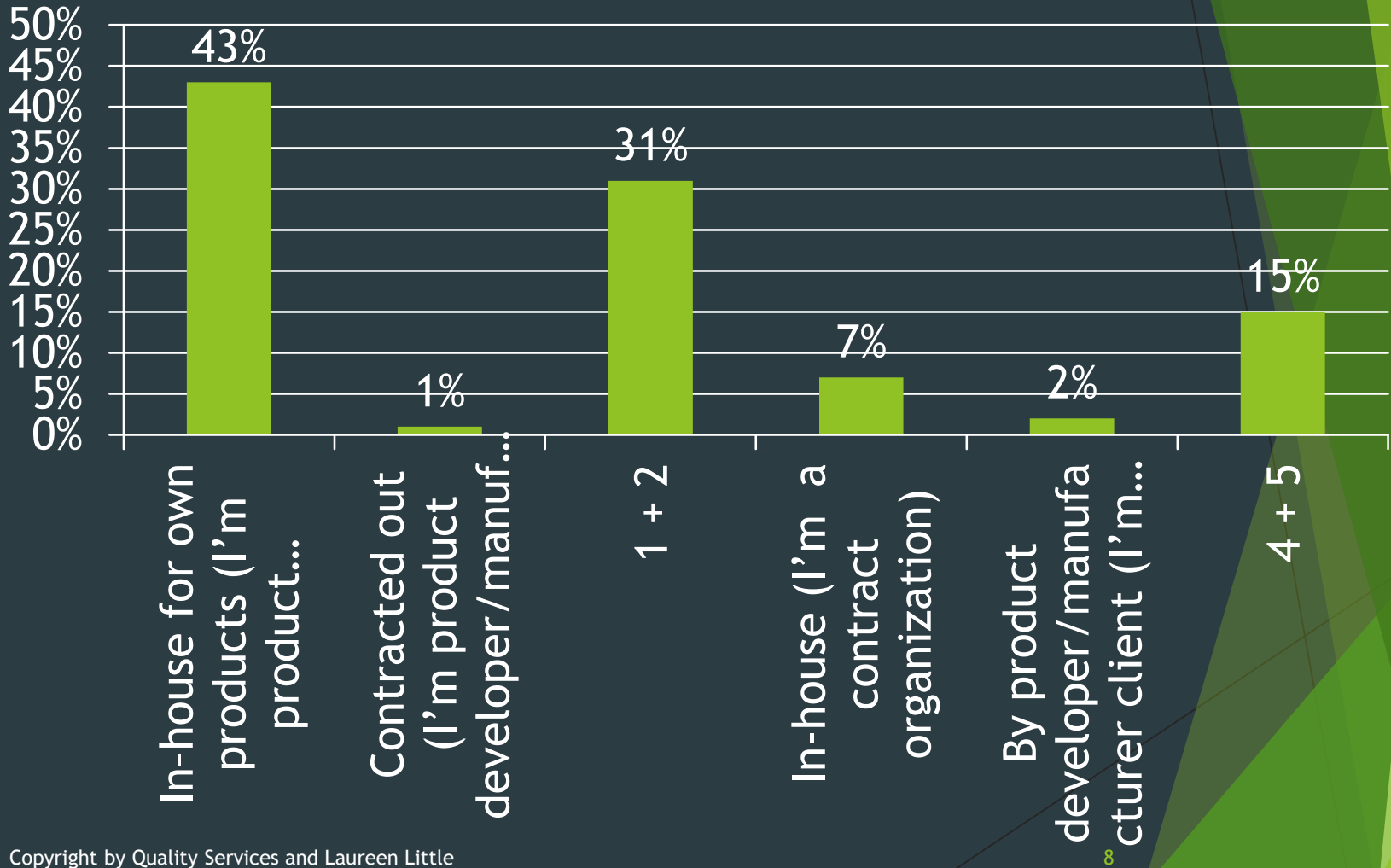
1. Contract Organization
2. Small Biopharmaceutical (< 50 employees)
3. Mid-Size Biopharmaceutical (50 - 300)
4. Global Pharmaceutical
5. Consultant
6. Research Institute
7. Other



# Where are the bioassays developed?

1. In-house for own products (I'm product developer/manufacturer)
2. Contracted out (I'm product developer/manufacturer)
3. 1 + 2
4. In-house (I'm a contract organization)
5. By product developer/manufacturer client (I'm a contract organization)
6. 4 + 5

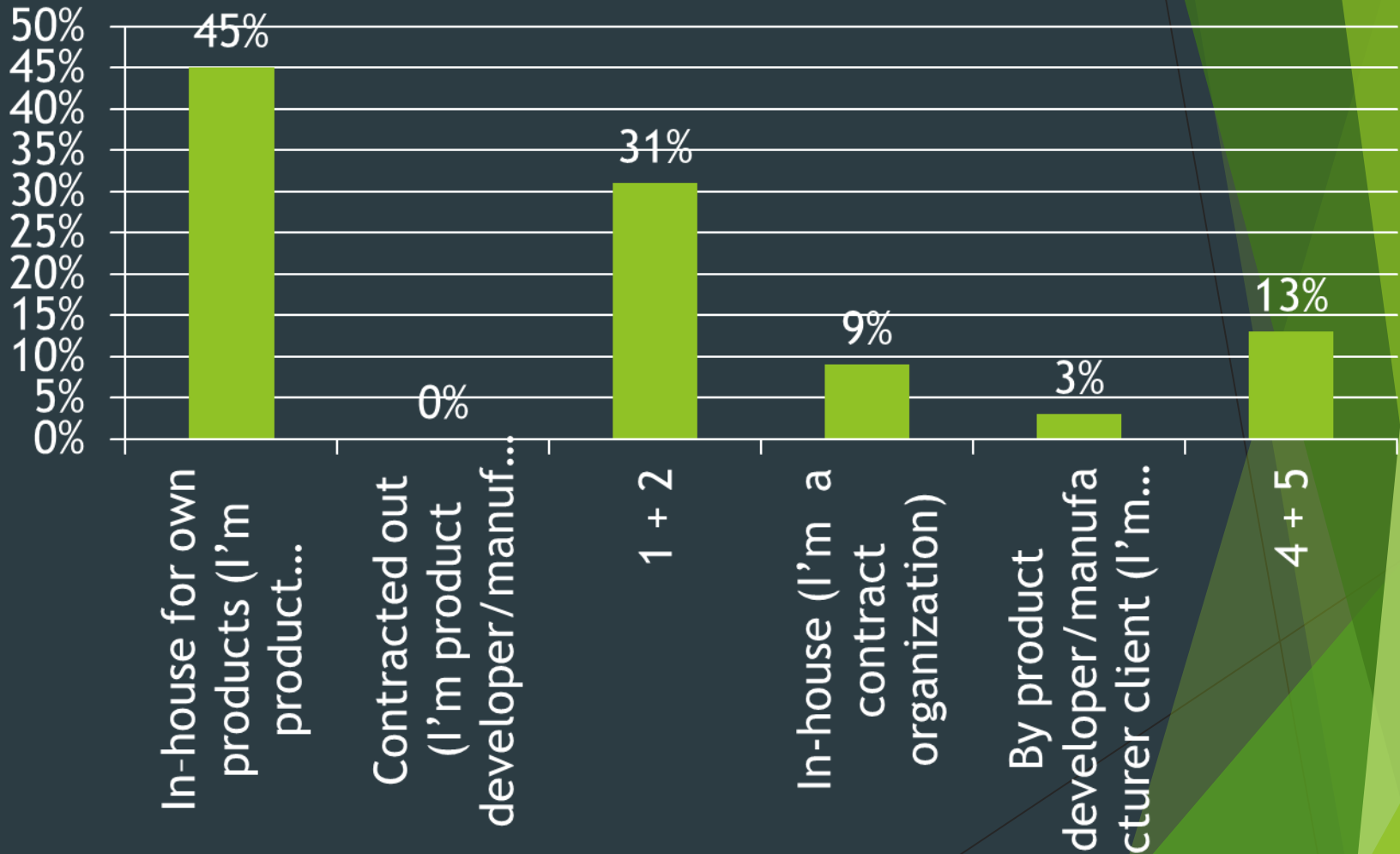
# Where?





# Where?

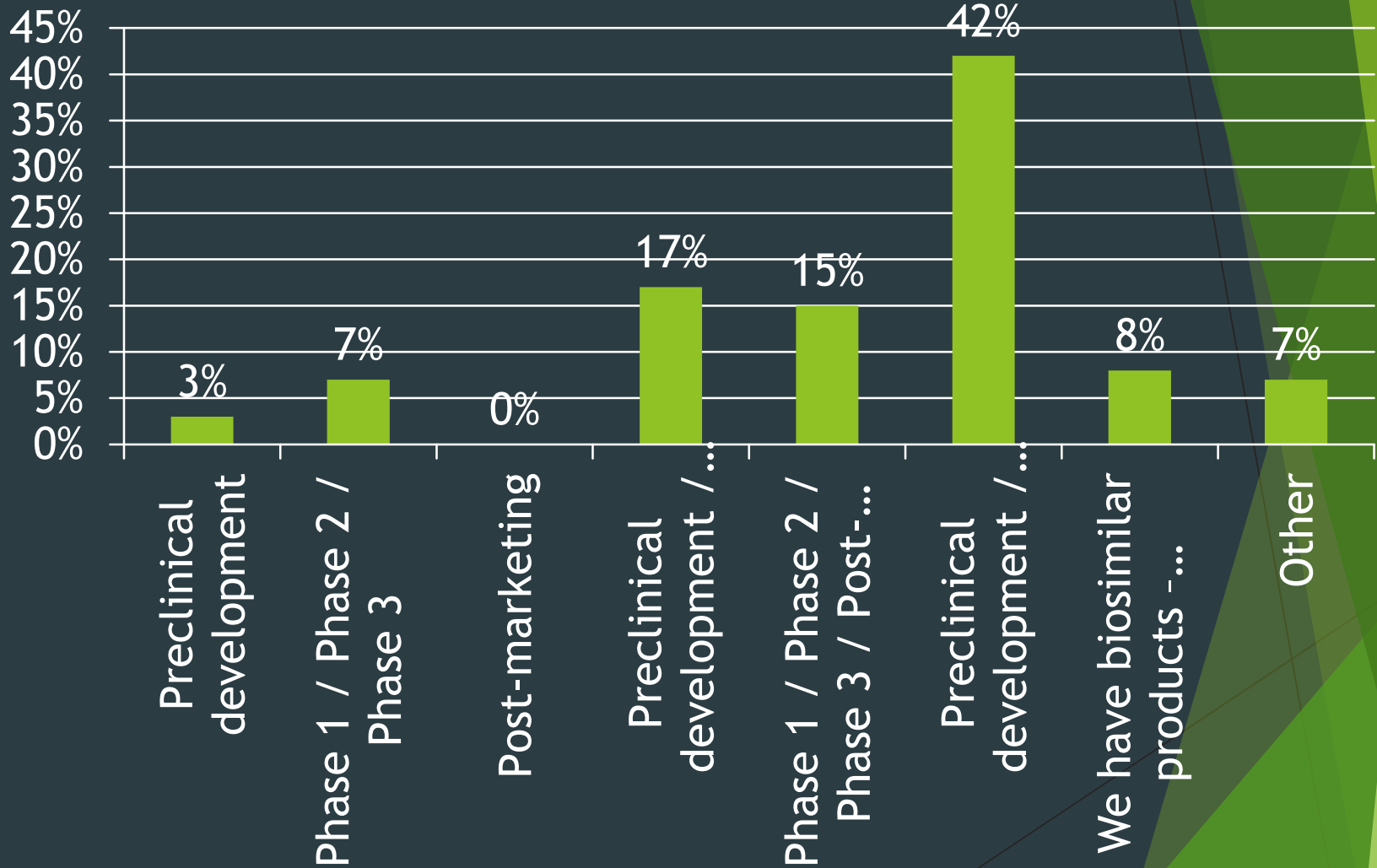
EU Responses



# Stages at which assay(s) used

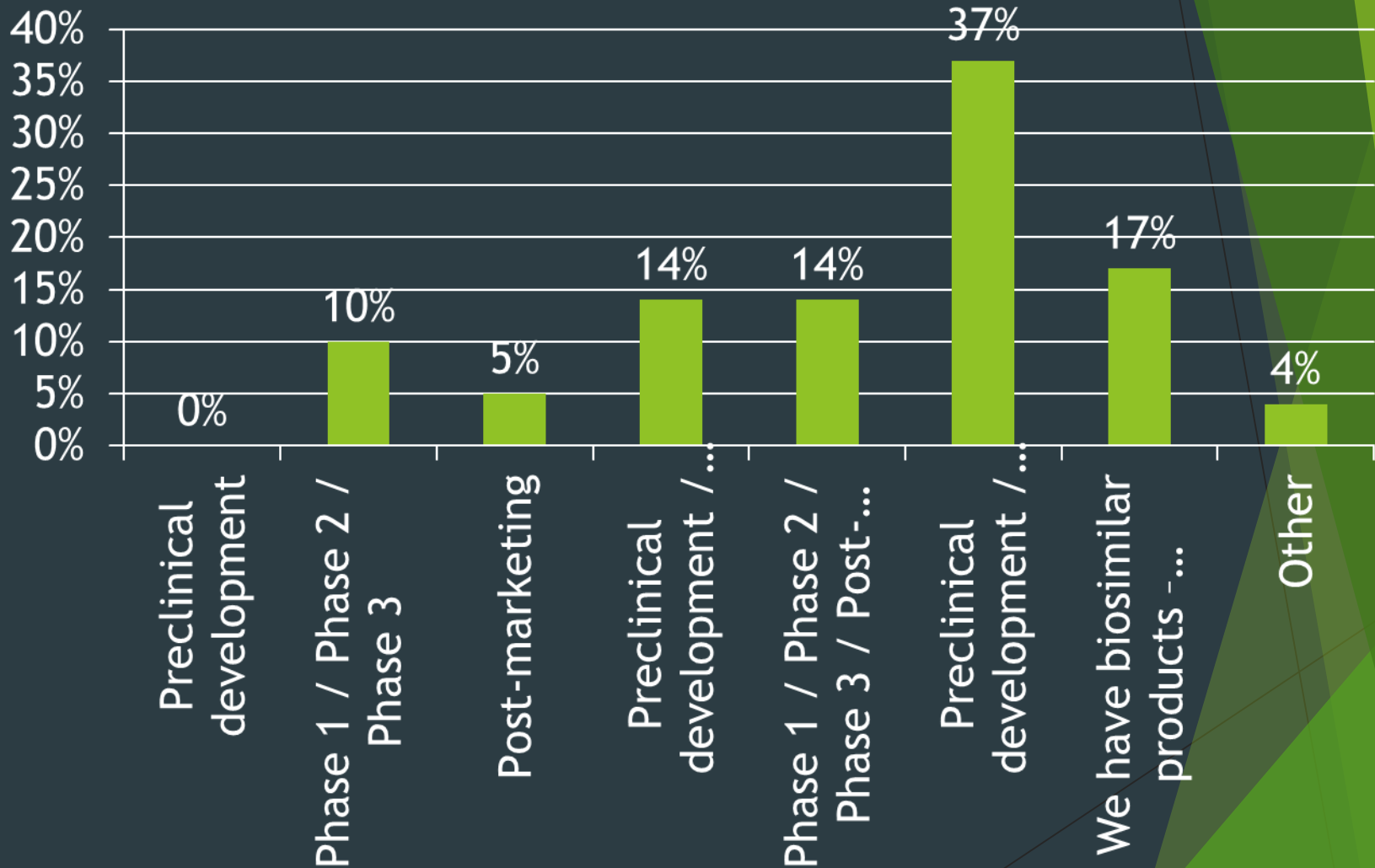
1. Preclinical development
2. Phase 1 / Phase 2 / Phase 3
3. Post-marketing
4. Preclinical development / Phase 1 / Phase 2 / Phase 3
5. Phase 1 / Phase 2 / Phase 3 / Post-marketing
6. Preclinical development / Phase 1 / Phase 2 / Phase 3 / Post-marketing
7. We have biosimilar products - therefore the above doesn't make sense
8. Other

# When?



# When?

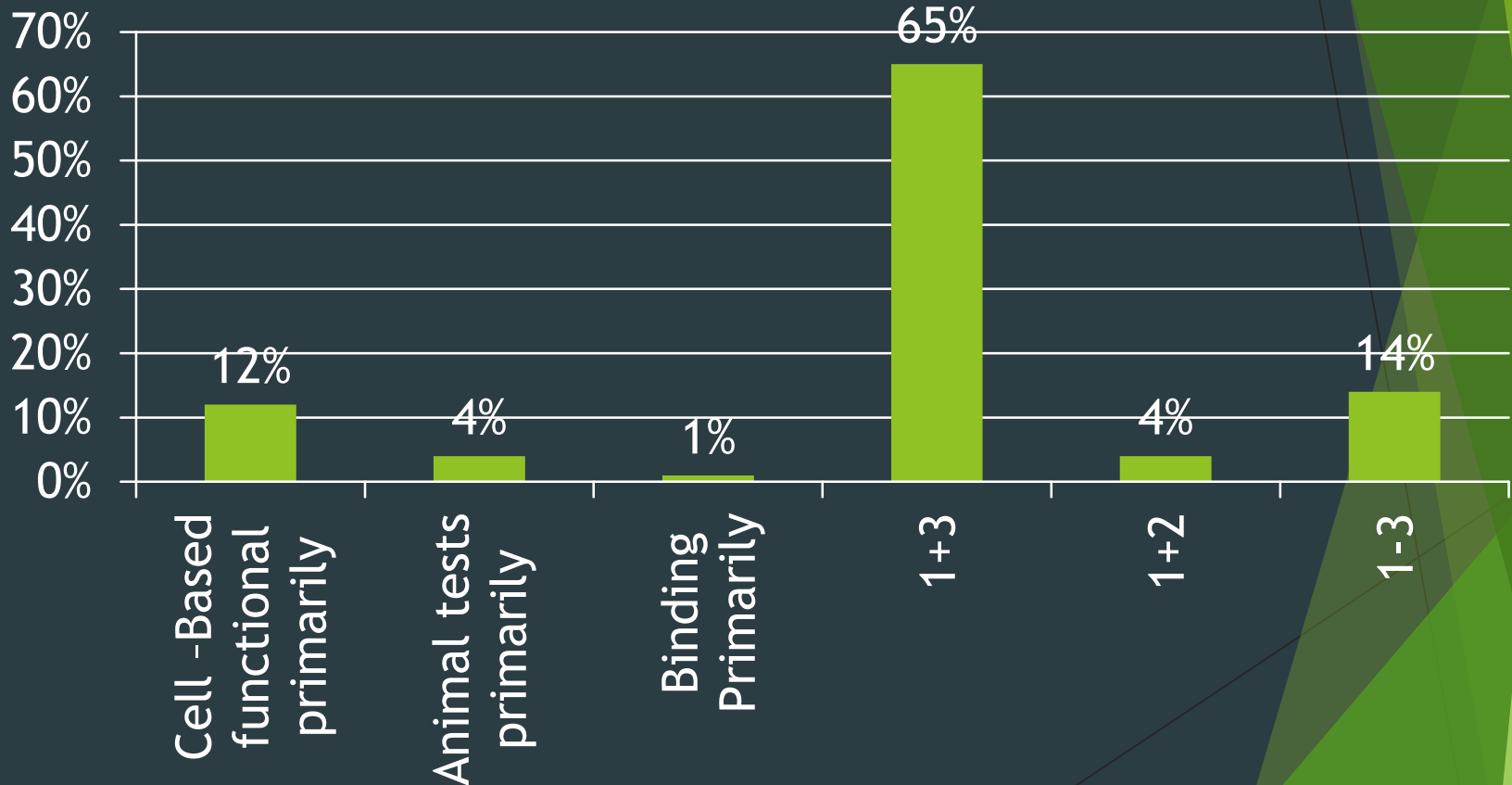
## EU Responses



# Functional or Ligand Binding?

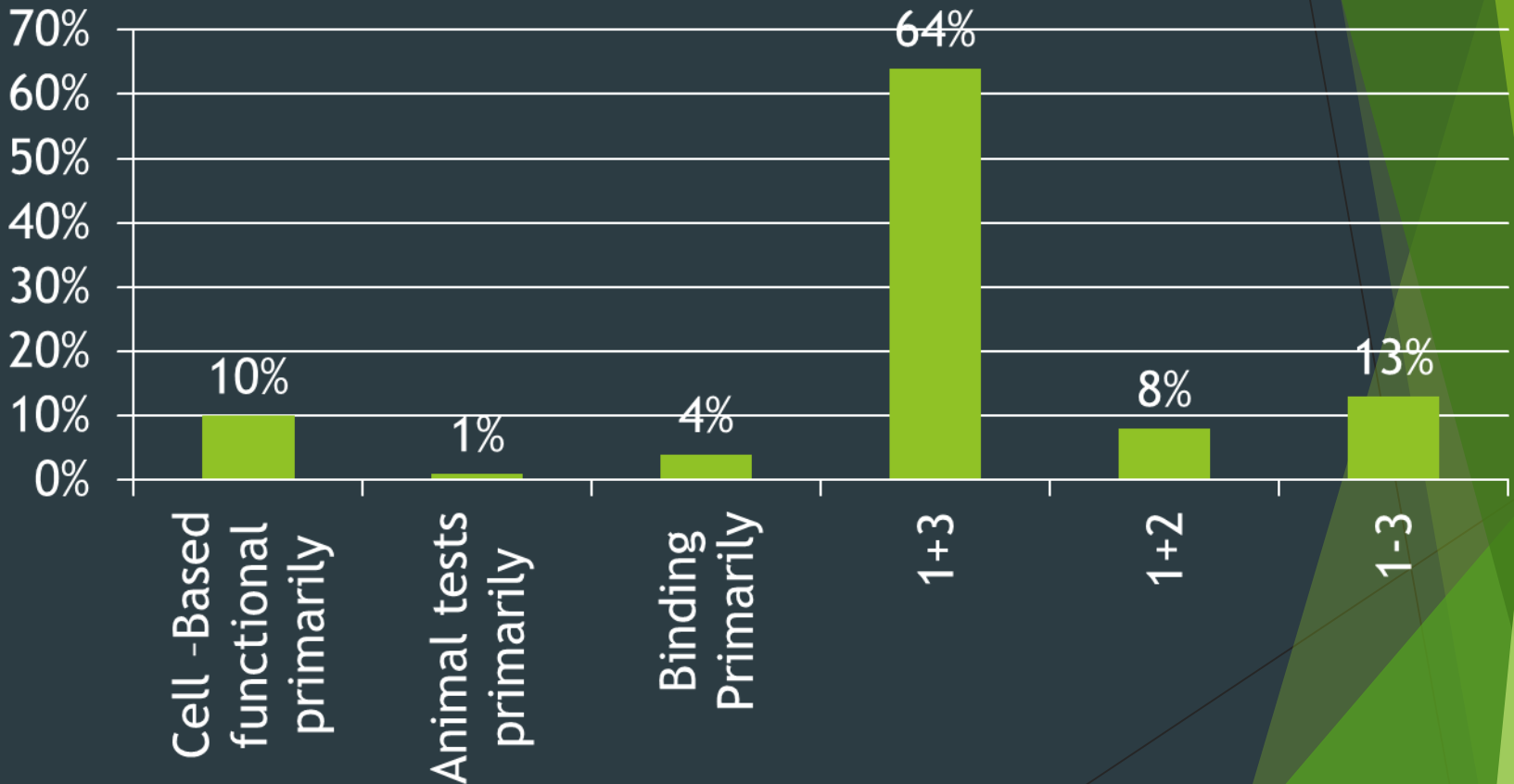
1. Cell -Based functional primarily
2. Animal tests primarily
3. Binding Primarily
4. 1+3
5. 1+2
6. 1-3

# Types of Assays



# Types of Assays

EU Response

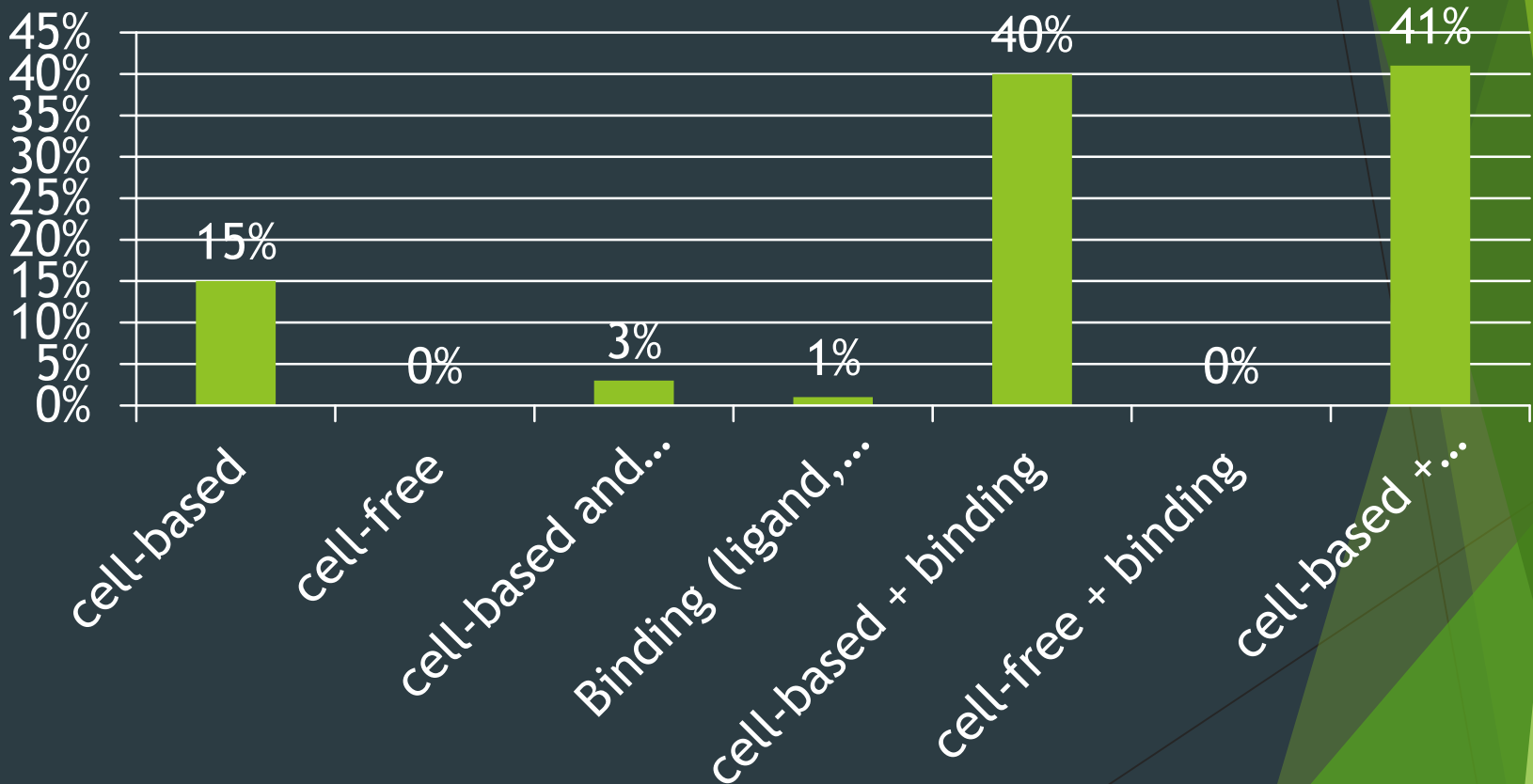


# Functional Assay types used

1. cell-based
2. cell-free
3. cell-based and functional cell-free
4. Binding (ligand, receptor, cofactor, ....)
5. cell-based + binding
6. cell-free + binding
7. cell-based + functional cell-free + binding

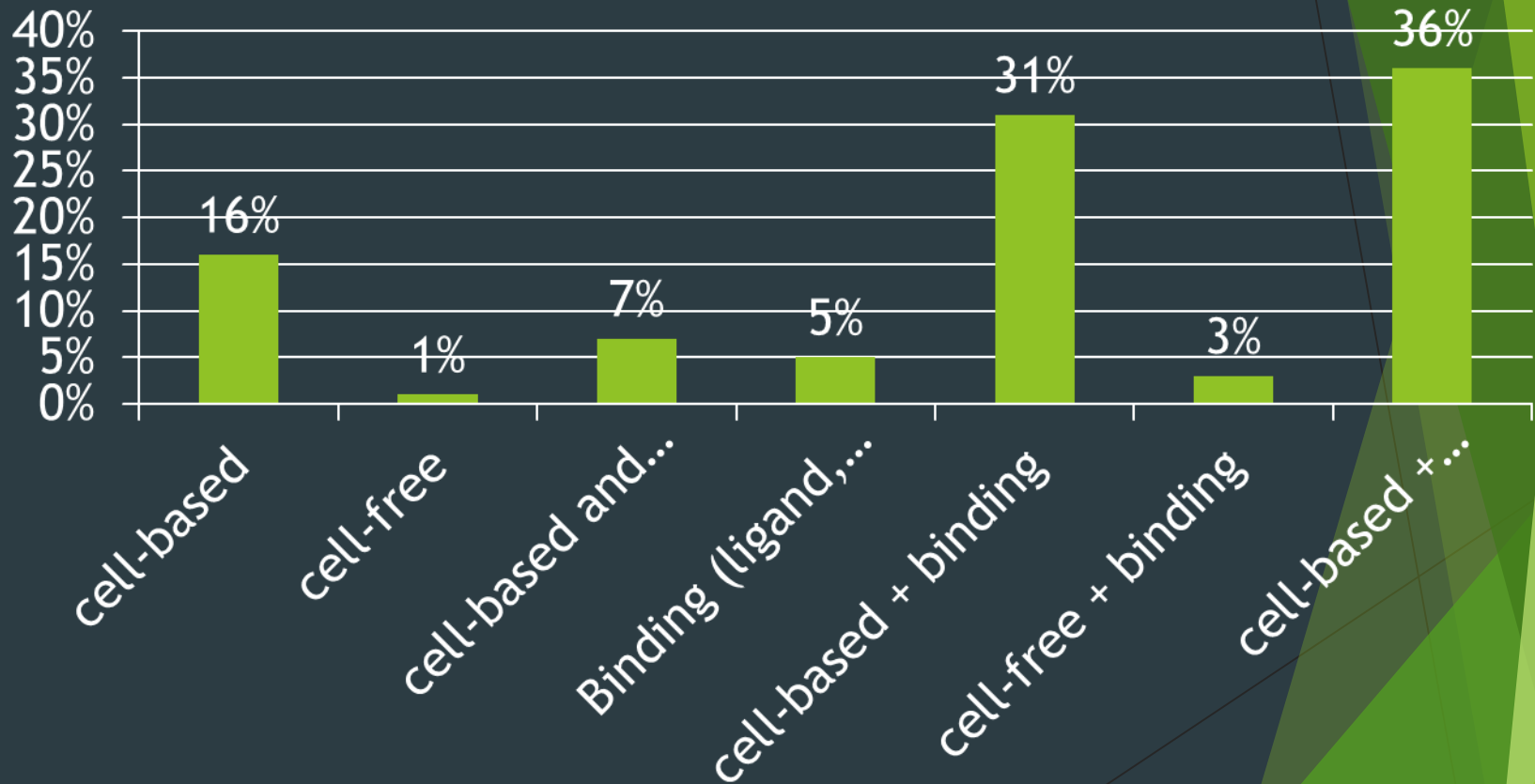


# Functional Assay types used



# Functional Assay Types

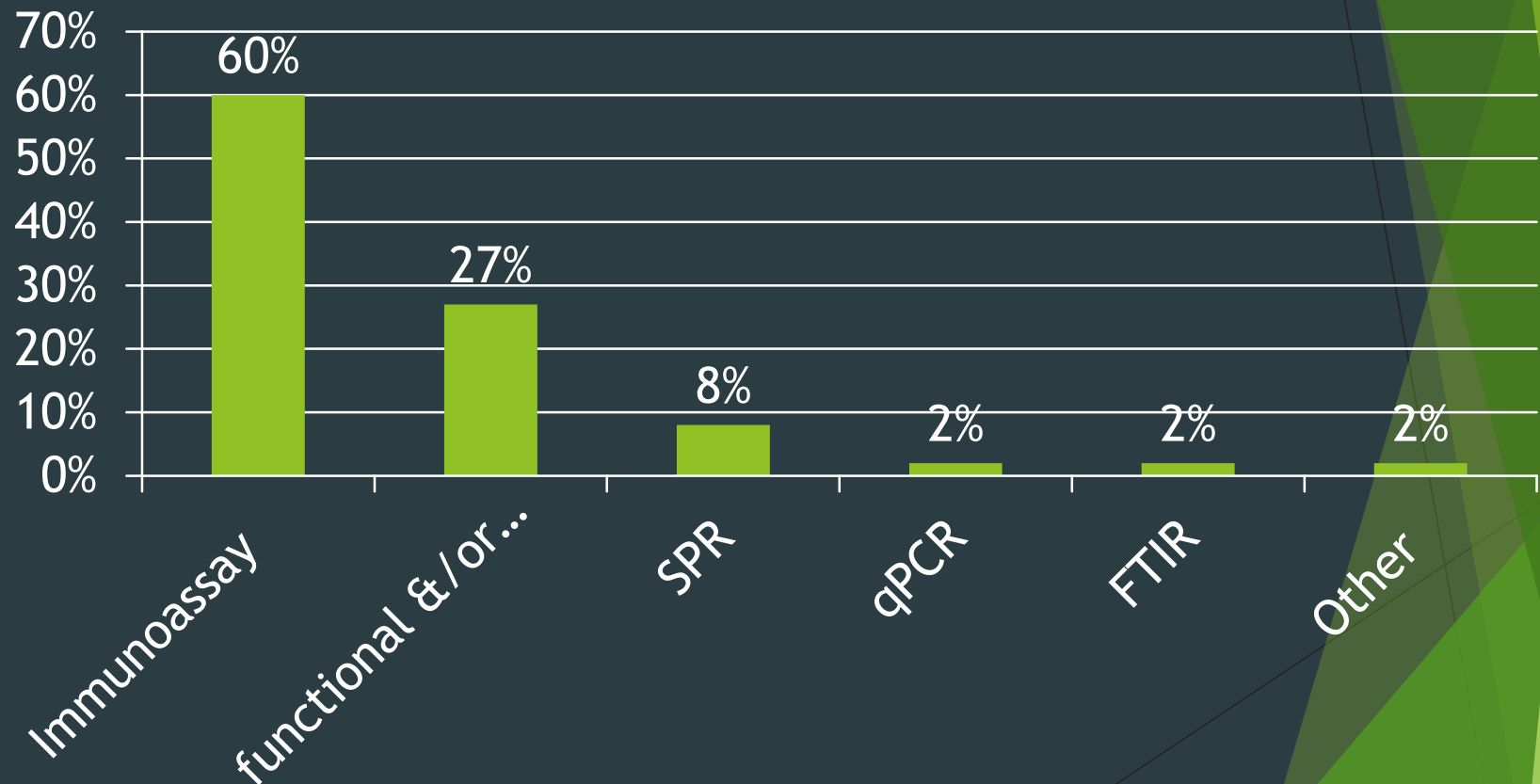
EU Response



# Binding Assay type (primarily)

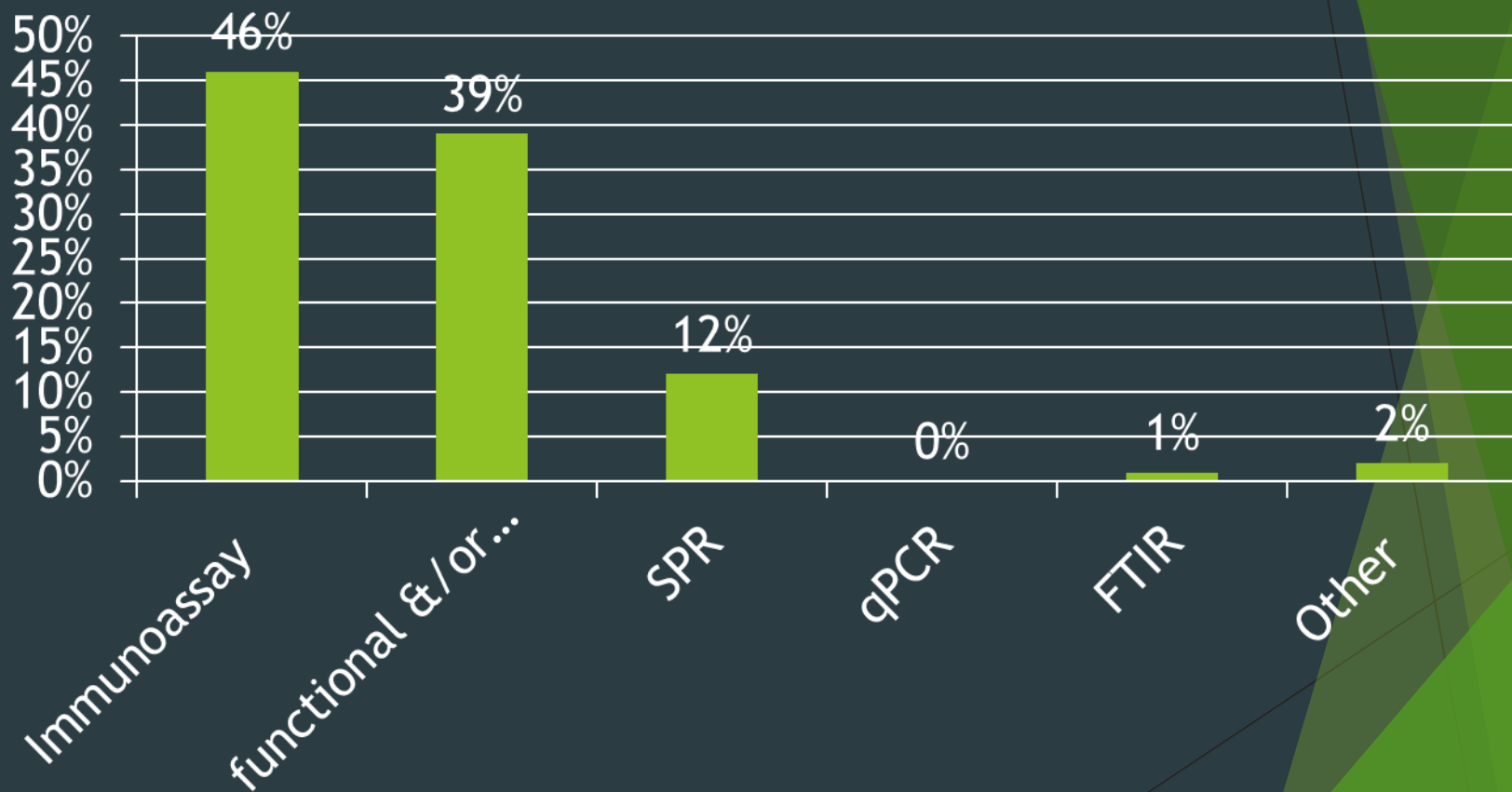
1. Immunoassay
2. functional &/or binding
3. SPR
4. qPCR
5. FTIR
6. Other

# Ligand Binding Assays



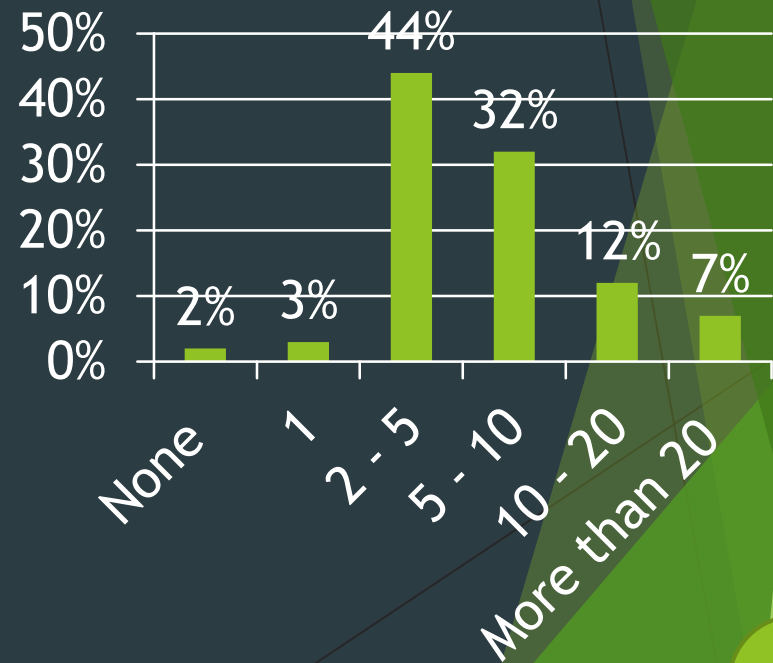
# Ligand Binding Assays

EU Responses



# How many bioassay systems do you run?

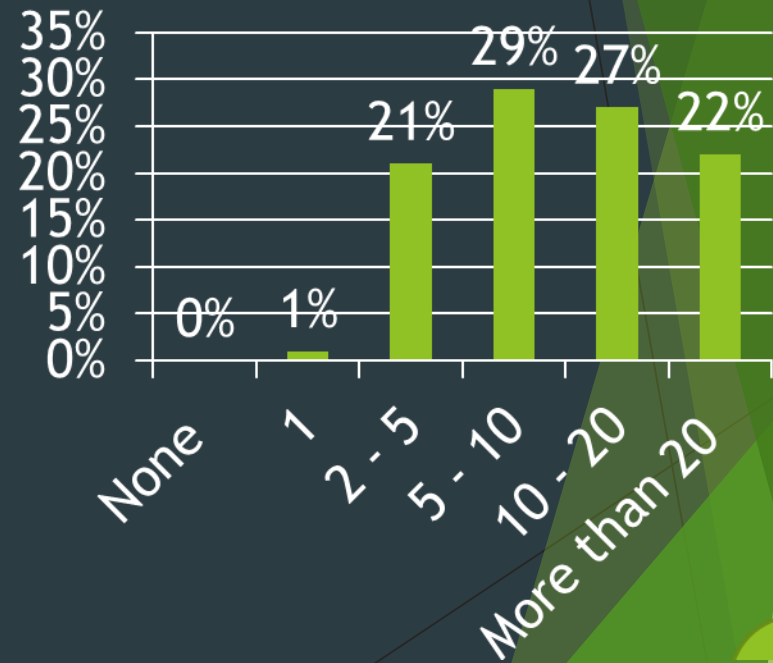
1. None
2. 1
3. 2 - 5
4. 5 - 10
5. 10 - 20
6. More than 20



# How many bioassay systems do you run?

1. None
2. 1
3. 2 - 5
4. 5 - 10
5. 10 - 20
6. More than 20

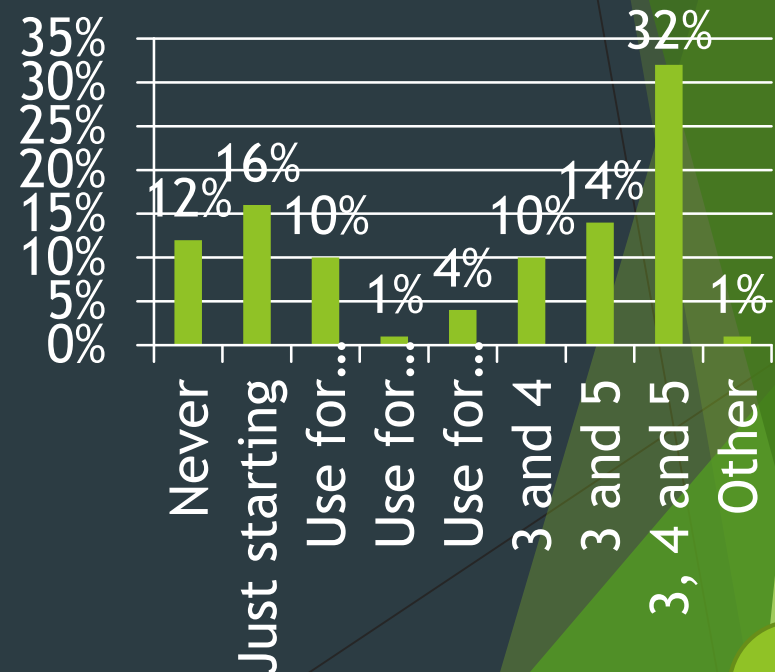
## EU Response



# DOE in your lab

What is your current use of DOE?

1. Never
2. Just starting
3. Use for robustness only
4. Use for trouble shooting
5. Use for component optimization
6. 3 and 4
7. 3 and 5
8. 3, 4 and 5
9. Other



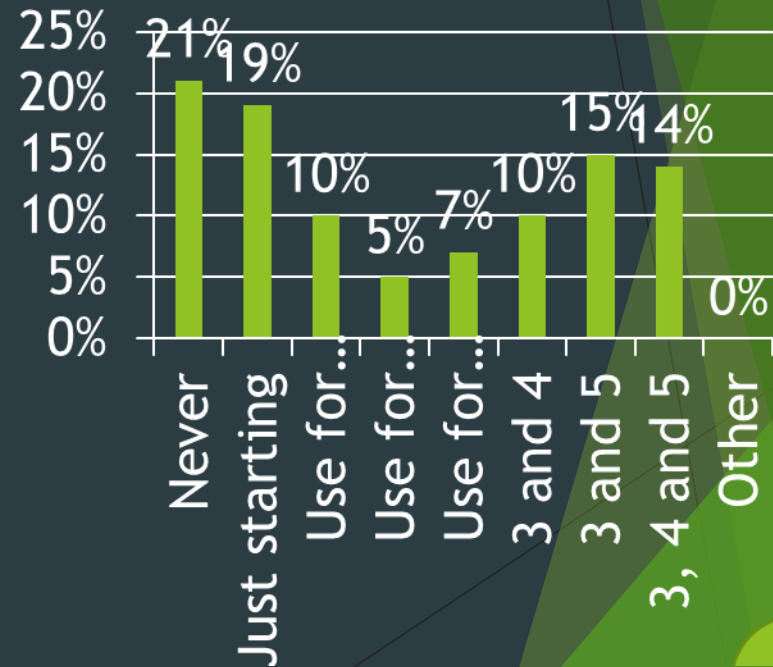


# DOE in your lab

## EU Response

What is your current use of DOE?

1. Never
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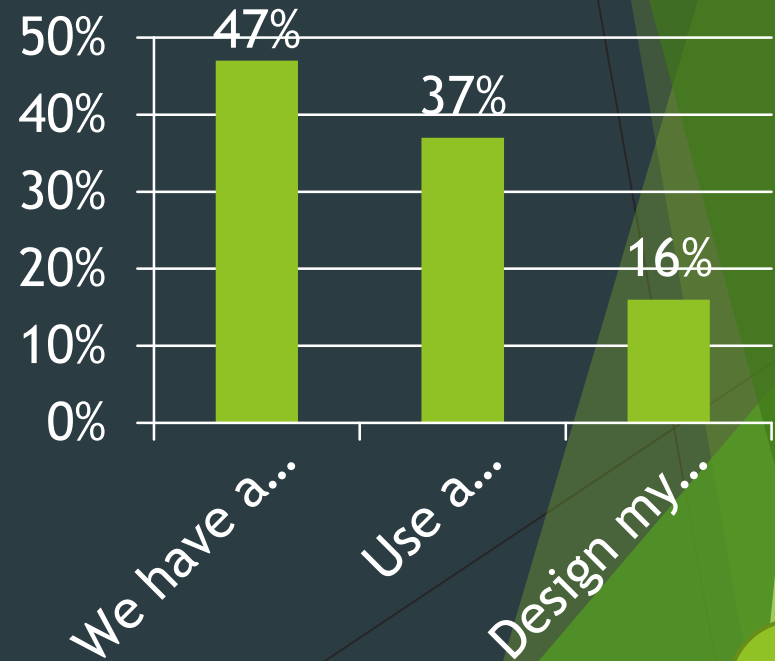
# Design of Experiments (DOE) in Bioassays

- ▶ DOE is a tool which can be used throughout the entire development cycle.
- ▶ It is best used sequentially (i.e. don't try to design *one* experiment to ask all your development questions).
- ▶ Current bioassay field uses DOE to determine robustness. While this is a fabulous tool - if it is your only use, then you are starting too late!

# Your Designs

How do you design you DOEs

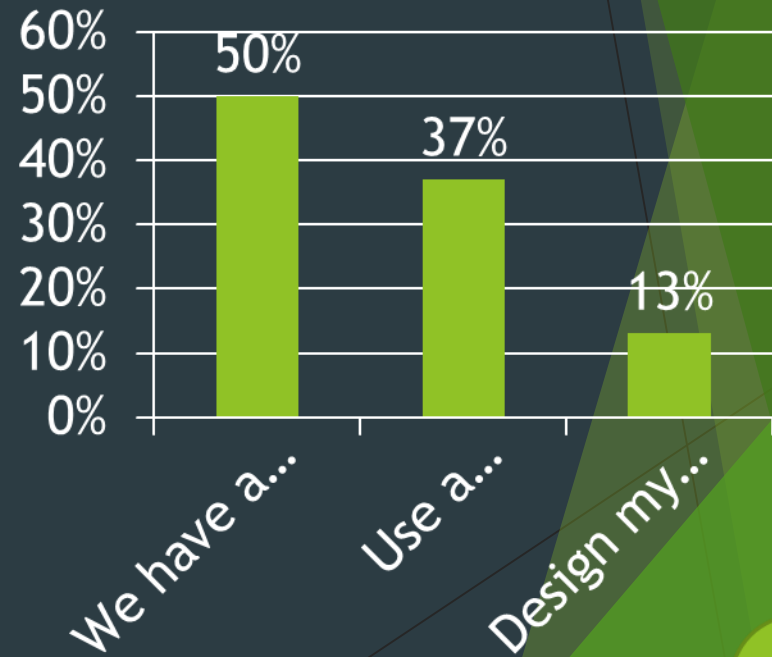
1. We have a statistician (either employed or consultant)
2. Use a software and design my own.
3. Design my own without software



# Your Designs EU Response

How do you design you DOEs

1. We have a statistician (either employed or consultant)
2. Use a software and design my own.
3. Design my own without software



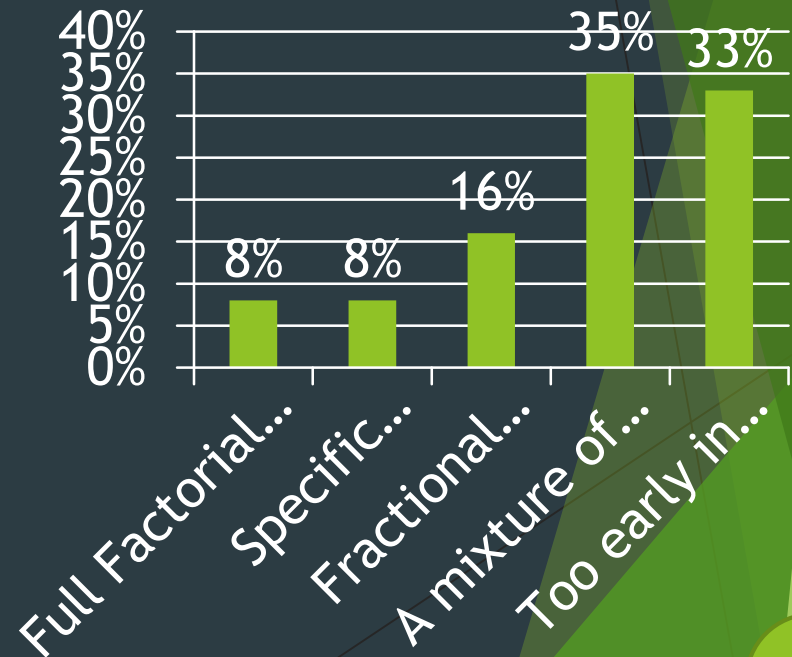
# Types of DoE

- ▶ Screening methodologies : which are designed to determine what factors are important.
  - ▶ Fractional Factorials
  - ▶ Specialized designs such as Taguchi (Plackett-Burman)
- ▶ Full Factorials: which are designed to determine the best conditions of the factors you know to be important
  - ▶ Most common one we see:  $2^3$  factorial

# Your Design (Continued)

What type of designs do you use?

1. Full Factorial only
2. Specific screening design only (such as a placket-burman)
3. Fractional Factorial
4. A mixture of the above
5. Too early in our use of DOE to be able to answer this question

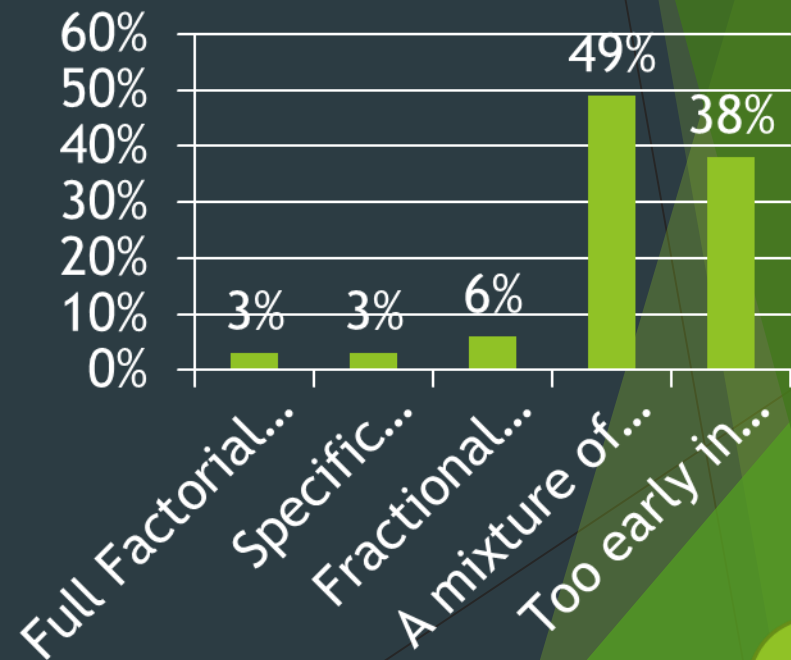


# Your Design (Continued)

EU Response

What type of designs do you use?

1. Full Factorial only
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5. Too early in our use of DOE to be able to answer this question



# Cell Culture Example

- ▶ This was a screening design - we were trying to optimize a component - the cell culture - we didn't know what was important.
- ▶ Choose 5-6 factors and design a Fractional Factorial.
- ▶ This can be done twice.
- ▶ We may find that most of the factors we think we should study - don't actually impact the method. Therefore it is smart to figure out which factors are critical and then study them.



# This is a Sequential DOE Approach

- ▶ Following is a sequential method development - using sequential DOE.
- ▶ The following example is a design to determine the best growing conditions for a cell-based potency assay. ***Why? It is the most crucial component for achieving low imprecision.***

# Choosing the Right Response

- ▶ Most of the DOEs that I have seen have not carefully thought through what should be the measured response.
- ▶ This is especially important if you are trying to optimize the assay by improving a specific characteristic or component of the assay.
- ▶ ***IN THIS EXAMPLE WE DID NOT HAVE DRUG PRESENT!! Since we were only trying to optimize the cells we only looked at a viability dye.***

# In Other Assay Systems I would have:

- ▶ Perhaps looked at specific receptor expression on the cell
- ▶ Looked at zero and high drug concentrations
- ▶ Perhaps other viability marker?
  
- ▶ But when optimizing components - you normally do not want to look at the entire assay.

# Additional Examples

## ▶ ELISA assays

- ▶ We were having a non-specific binding problem. We used sequential DOE to identify blocking reagents and procedures to essentially eliminate this. Our read out was average of 5 blank samples and 5 high samples. We calculated results for both Blank average and Signal/Noise ( $Z'$  factor)

## ▶ Cell based assay (biomarker)

- ▶ We were having dilutional linearity problems. We used sequential DOE. Here we used 4 single point dilutions of several patient samples and looked at “average” relative bias numbers. We found a sample diluent which completely solved the problem. (This took 5 placket-burman runs, followed by 2 full factorials)

# Back to example at Hand: Some Interesting Things about this Design

- ▶ It was chosen to do this in 16 plates (based upon the tissue culture analyst that could handle 8 plates)
  - ▶ Note the analysts actually informed me they could easily handle 16 plates per day. I assumed that in any type of DOE throughput drops by 50% because of the complexity of the individual assay runs.
- ▶ We chose 6 variables.

# First Select the Factors and Levels

Name	Units	Type	Low	High
Passage #	passage	Factor	<P10	>P26
Seeding Density	cells per cm2	Factor	2000	4000
Tryp concentration	ml	Factor	2	4
Tryp Incub	Minutes	Factor	5	10
FBS lots	Lot	Factor	B1	B2
Day Feeding	times per week	Factor	1	2
R1	%CV*	Response		
R2	Bowl Ratio**	Response		

# Initial Screen from StatEase

	Number of Factors												
	2	3	4	5	6	7	8	9	10	11	12	13	14
4	$2^2$	$2^{3-1}$ III											
8		$2^3$	$2^{4-1}$ IV	$2^{5-2}$ III	$2^{6-3}$ III	$2^{7-4}$ III							
16			$2^4$	$2^{5-1}$ V	$2^{6-2}$ IV	$2^{7-3}$ IV	$2^{8-4}$ IV	$2^{9-5}$ III	$2^{10-6}$ III	$2^{11-7}$ III	$2^{12-8}$ III	$2^{13-9}$ III	$2^{14-10}$ III
32				$2^5$	$2^{6-1}$ VI	$2^{7-2}$ IV	$2^{8-3}$ IV	$2^{9-4}$ IV	$2^{10-5}$ IV	$2^{11-6}$ IV	$2^{12-7}$ IV	$2^{13-8}$ IV	$2^{14-9}$ IV
64					$2^6$	$2^{7-1}$ VII	$2^{8-2}$ V	$2^{9-3}$ IV	$2^{10-4}$ IV	$2^{11-5}$ IV	$2^{12-6}$ IV	$2^{13-7}$ IV	$2^{14-8}$ IV
128						$2^7$	$2^{8-1}$ VIII	$2^{9-2}$ VI	$2^{10-3}$ V	$2^{11-4}$ V	$2^{12-5}$ IV	$2^{13-6}$ IV	$2^{14-7}$ IV
256							$2^8$	$2^{9-1}$ IX	$2^{10-2}$ VI	$2^{11-3}$ VI	$2^{12-4}$ VI	$2^{13-5}$ V	$2^{14-6}$ V
512								$2^9$	$2^{10-1}$ X	$2^{11-2}$ VII	$2^{12-3}$ VI	$2^{13-4}$ VI	$2^{14-5}$ VI

↑ Number of runs

This gave us a  $2^{6-2}$  Fractional Factorial.  
This is a level 4 resolution.  
What does this mean?

Resolution	Ability
II	Not useful: main effects are confounded with other main effects
III	Estimate main effects, but these may be confounded with two-factor interactions
IV	Estimate main effects unconfounded by two-factor interactions Estimate two-factor interaction effects, but these may be confounded with other two-factor interactions
V	Estimate main effects unconfounded by three-factor (or less) interactions Estimate two-factor interaction effects unconfounded by two-factor interactions Estimate three-factor interaction effects, but these may be confounded with other two-factor interactions
VI	Estimate main effects unconfounded by four-factor (or less) interactions Estimate two-factor interaction effects unconfounded by three-factor (or less) interactions Estimate three-factor interaction effects, but these may be confounded with other three-factor interactions



# What is the Response?

- ▶ This is component optimization
  - not assay optimization.
- ▶ Therefore, what are the characteristics we would like to see?
  - ▶ Well-to-well consistency of growth (This can be measured by an Alamar Blue dye, cell-titer glo, whatever viability assay you have - then reported out as an average and %CV.)
  - ▶ Lack of systematic bias: Experience tells us that the bowl ratio is the most common growth pattern: Therefore, let's take Avg OD outer / Avg OD inner

# Run Design

Std	Run	Passage	Seeding	[Tryp]	Try Incub	FBS	Feeding	R1	R2
		P#	cells per cm2	ml	Minutes	Lot	X /week	%CV	Bowl Ratio
1	1	<P6	5000	2	10	B2	2		
9	2	<P6	5000	4	10	B1	1		
2	3	<P6	5000	2	10	B2	2		
10	4	<P6	5000	4	10	B1	1		
13	5	<P6	10000	4	5	B1	2		
5	6	<P6	10000	2	5	B2	1		
14	7	<P6	10000	4	5	B1	2		
6	8	<P6	10000	2	5	B2	1		
11	9	>P12	5000	4	5	B2	1		
3	10	>P12	5000	2	5	B1	2		
12	11	>P12	5000	4	5	B2	1		
4	12	>P12	5000	2	5	B1	2		
7	13	>P12	10000	2	10	B1	1		
15	14	>P12	10000	4	10	B2	2		
8	15	>P12	10000	2	10	B1	1		
16	16	>P12	10000	4	10	B2	2		

# Results

- ▶ All of the results had a serious positional problem.
- ▶ None of the factors studied had an impact.
- ▶ Did a second round of experimentation looking at more of the technique issues.

Name	Units	Type	Low	High
Initial mixing of cells	Y/N	Factor	Simple Inversion	10 x inversion
Mixing prior to dispensing	Y/N	Factor	up/down 2 x in pipette	on rotary
pipette type used	pipette type	Factor	12 well	96-well
Temperature (media)	degrees	Factor	25	37
pipette tips	brand	Factor	B1	B2
Trypsinzation	time (minutes)	Factor	5	15
R1	%CV*	Response		
R2	1st row vs last row	Response		

C:\Users\Public\Documents\DX8 data\MyDesign.dxp - Design-Expert 8.0.7.1

File Edit View Display Options Design Tools Help Tips

Combined

Mixture

Response Surface

Factorial

2-Level Factorial

Min-Run Res V

**Min-Run Res IV**

Irregular Fraction

General Factorial

Optimal

Plackett-Burman

Taguchi OA

## Minimum-Run Equireplicated Res IV Design

Design for 5 to 50 factors where each factor is varied over only 2 levels. Resolution IV designs will allow estimation of main effects. Two-factor interactions will be aliased with other two-factor and higher interactions. Good designs to reduce the number of runs if interactions are unlikely

Factors:  (5 to 50)       Horizontal  
 Vertical

	Name	Units	Type	Low	High
A [Categorical]	Initial Mix		Categorical	Simple	10x
B [Categorical]	Mix prior to dis		Categorical	up/down 2x	On Rotary
C [Categorical]	Pipette		Categorical	12 well	96 well
D [Numeric]	Temperature	degrees	Numeric	25	37
E [Categorical]	Pipette Tips		Categorical	B1	B2
F [Numeric]	Trypsinization	minutes	Numeric	5	15

Suggestion: Only use "Min-Run Res IV" designs if you cannot afford the runs required by an orthogonal "2-Level Factorial" design. Consider using the orthogonal resolution IV (yellow) designs under "2-Level Factorial."

Minimum runs plus 2  
 Minimum runs

Center points:       14 Runs

# Original Run Lay Out Selected by Software

Select	Std	Run	Factor 1 A:Initial Mix	Factor 2 B:Mix prior to...	Factor 3 C:Pipette	Factor 4 D:Temperature degrees	Factor 5 E:Pipette Tips	Factor 6 F:Trypsinizat... minutes	Response 1 R1 %CV	Response 2 R2 Ratio
4		1	Simple	up/down 2x	96 well	25.00	B2	5.00		
	7	2	Simple	On Rotary	12 well	25.00	B2	15.00		
	11	3	10x	On Rotary	12 well	37.00	B1	15.00		
	2	4	Simple	up/down 2x	12 well	37.00	B1	5.00		
	14	5	10x	up/down 2x	96 well	37.00	B1	5.00		
	13	6	Simple	On Rotary	12 well	25.00	B1	5.00		
	9	7	10x	On Rotary	96 well	25.00	B2	15.00		
	1	8	Simple	On Rotary	12 well	37.00	B2	5.00		
	3	9	Simple	On Rotary	96 well	37.00	B1	15.00		
	10	10	10x	up/down 2x	12 well	25.00	B2	5.00		
	12	11	10x	On Rotary	96 well	37.00	B2	5.00		
	6	12	10x	up/down 2x	96 well	37.00	B2	15.00		
	8	13	10x	up/down 2x	96 well	25.00	B1	15.00		
	5	14	Simple	up/down 2x	12 well	25.00	B1	15.00		

**Design Tool** x

- Design Layout
- Run Sheet
- Column Info Sheet
- Pop-Out View

Problem is that having mixing procedures intertwined is procedurally difficult.

# Sorted by Factor 1

File Edit View Display Options Design Tools Help Tips										
Select	Std	Run	Factor 1 A:Initial Mix	Factor 2 B:Mix prior to...	Factor 3 C:Pipette	Factor 4 D:Temperature degrees	Factor 5 E:Pipette Tips	Factor 6 F:Trypsinizat... minutes	Response 1 R1 %CV	Response 2 R2 Ratio
4		1	Simple	up/down 2x	96 well	25.00	B2	5.00		
	7	2	Simple	On Rotary	12 well	25.00	B2	15.00		
	2	4	Simple	up/down 2x	12 well	37.00	B1	5.00		
	13	6	Simple	On Rotary	12 well	25.00	B1	5.00		
	1	8	Simple	On Rotary	12 well	37.00	B2	5.00		
	3	9	Simple	On Rotary	96 well	37.00	B1	15.00		
	5	14	Simple	up/down 2x	12 well	25.00	B1	15.00		
	11	3	10x	On Rotary	12 well	37.00	B1	15.00		
	14	5	10x	up/down 2x	96 well	37.00	B1	5.00		
	9	7	10x	On Rotary	96 well	25.00	B2	15.00		
	10	10	10x	up/down 2x	12 well	25.00	B2	5.00		
	12	11	10x	On Rotary	96 well	37.00	B2	5.00		
	6	12	10x	up/down 2x	96 well	37.00	B2	15.00		
	8	13	10x	up/down 2x	96 well	25.00	B1	15.00		

Still had a practical problem: Next sorted by Factor Two

# Final Run Lay Out

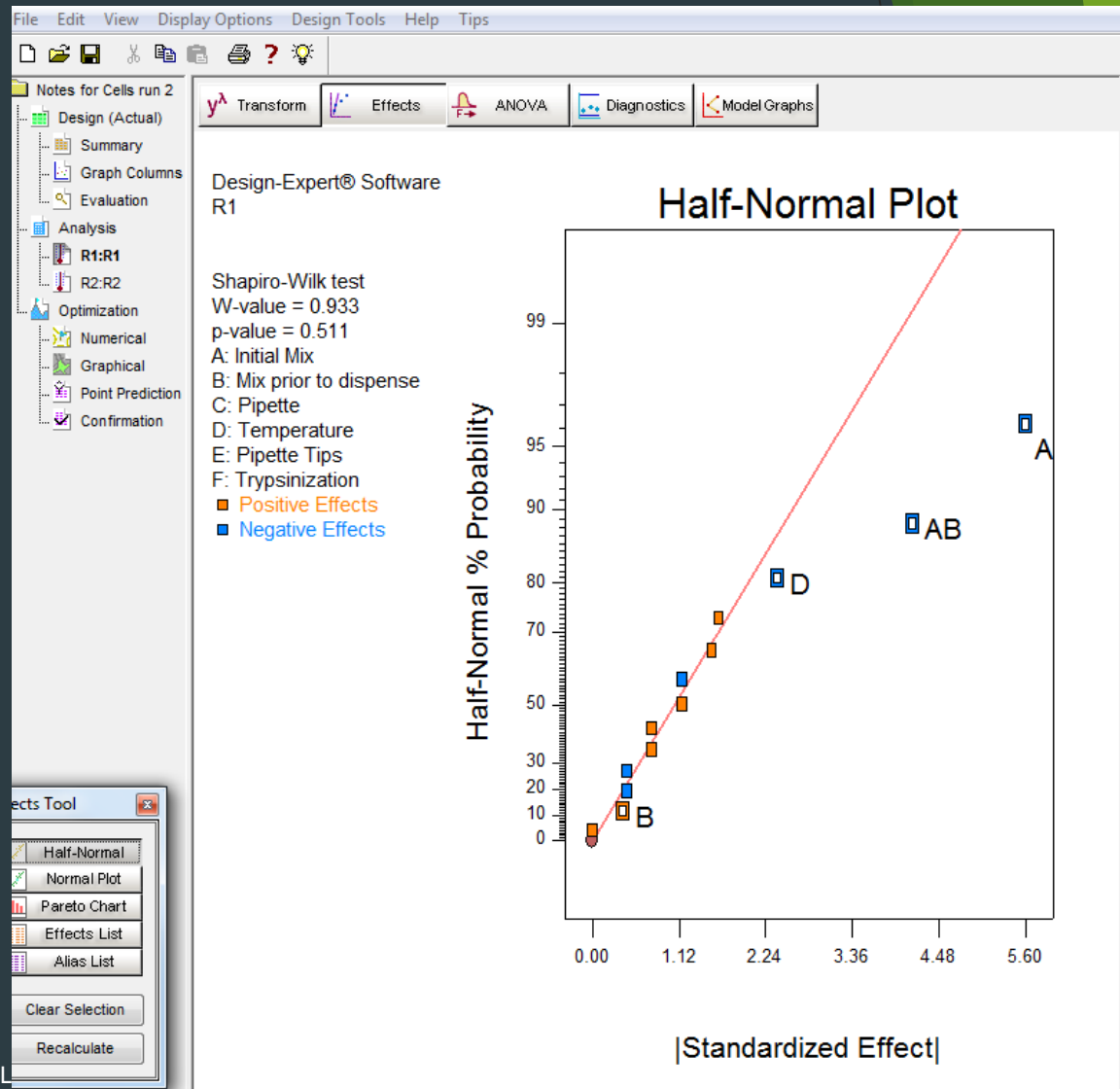
File Edit View Display Options Design Tools Help Tips										
Select	Std	Run	Factor 1 A:Initial Mix	Factor 2 B:Mix prior to...	Factor 3 C:Pipette	Factor 4 D:Temperature degrees	Factor 5 E:Pipette Tips	Factor 6 F:Trypsinizat... minutes	Response 1 R1 %CV	Response 2 R2 Ratio
4	1		Simple	up/down 2x	96 well	25.00	B2	5.00		
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	10	10	10x	up/down 2x	12 well	25.00	B2	5.00		
	6	12	10x	up/down 2x	96 well	37.00	B2	15.00		
	8	13	10x	up/down 2x	96 well	25.00	B1	15.00		
	7	2	Simple	On Rotary	12 well	25.00	B2	15.00		
	13	6	Simple	On Rotary	12 well	25.00	B1	5.00		
	1	8	Simple	On Rotary	12 well	37.00	B2	5.00		
	3	9	Simple	On Rotary	96 well	37.00	B1	15.00		
	11	3	10x	On Rotary	12 well	37.00	B1	15.00		
	9	7	10x	On Rotary	96 well	25.00	B2	15.00		
	12	11	10x	On Rotary	96 well	37.00	B2	5.00		



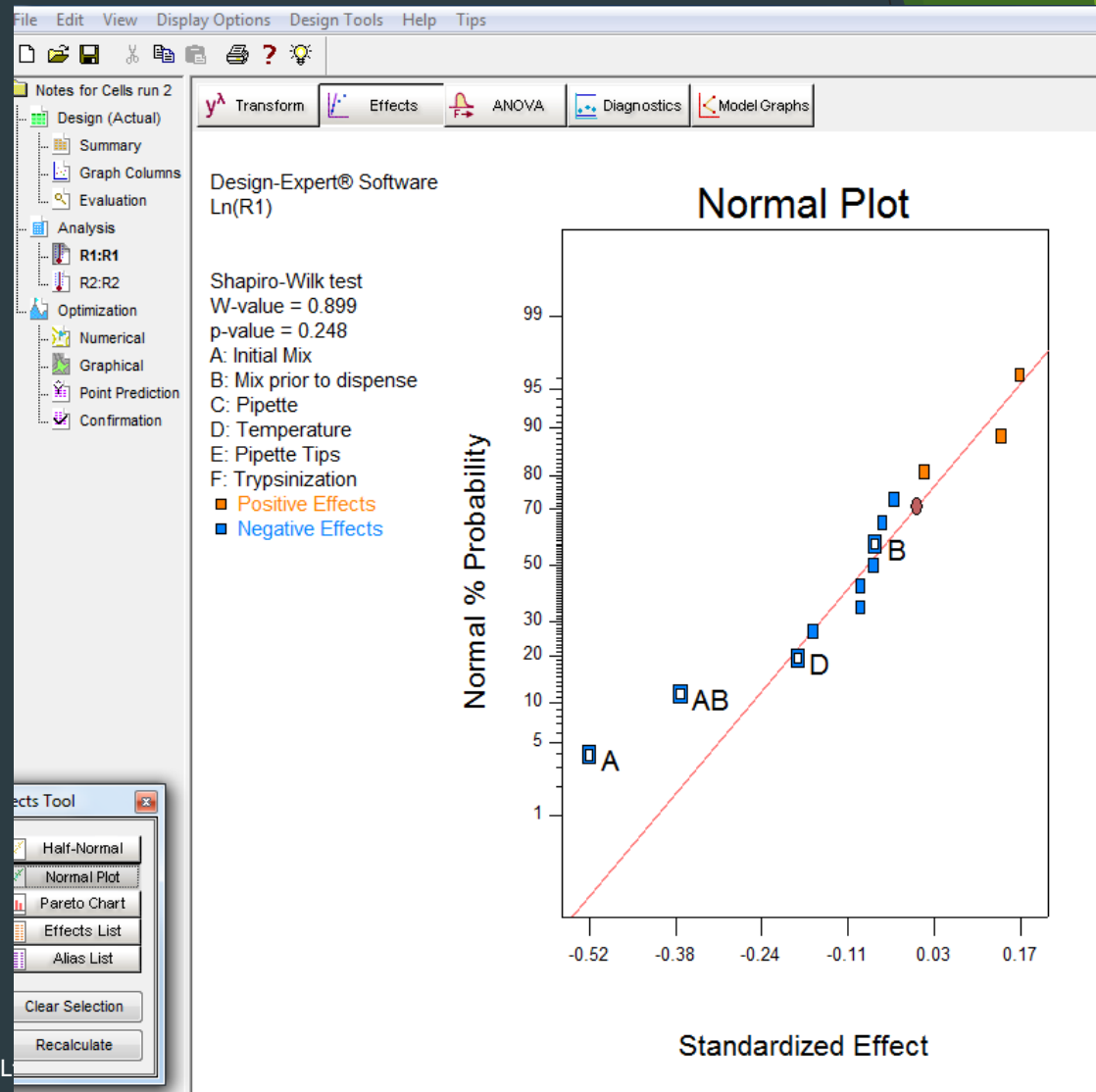
# Insert Data for Analysis

File Edit View Display Options Design Tools Help Tips											
Notes for Cells run 2											
Design (Actual)											
<ul style="list-style-type: none"> <li>Summary</li> <li>Graph Columns</li> <li>Evaluation</li> <li>Analysis                             <ul style="list-style-type: none"> <li>R1:R1 (Analyze)</li> <li>R2:R2</li> </ul> </li> <li>Optimization                             <ul style="list-style-type: none"> <li>Numerical</li> <li>Graphical</li> <li>Point Prediction</li> <li>Confirmation</li> </ul> </li> </ul>											
Select	Std	Run	Factor 1 A:Initial Mix	Factor 2 B:Mix prior to...	Factor 3 C:Pipette	Factor 4 D:Temperature degrees	Factor 5 E:Pipette Tips	Factor 6 F:Trypsinizat... minutes	Response 1 R1 %CV	Response 2 R2 Ratio	
4		1	Simple	up/down 2x	96 well	25.00	B2	5.00	12	1.5	
	7	2	Simple	On Rotary	12 well	25.00	B2	15.00	18	1.6	
	11	3	10x	On Rotary	12 well	37.00	B1	15.00	8	1.1	
	2	4	Simple	up/down 2x	12 well	37.00	B1	5.00	11	1.4	
	14	5	10x	up/down 2x	96 well	37.00	B1	5.00	7	1.2	
	13	6	Simple	On Rotary	12 well	25.00	B1	5.00	19	1.8	
	9	7	10x	On Rotary	96 well	25.00	B2	15.00	6	0.9	
	1	8	Simple	On Rotary	12 well	37.00	B2	5.00	12	1.4	
	3	9	Simple	On Rotary	96 well	37.00	B1	15.00	18	1.6	
	10	10	10x	up/down 2x	12 well	25.00	B2	5.00	12	1.3	
	12	11	10x	On Rotary	96 well	37.00	B2	5.00	6	1	
	6	12	10x	up/down 2x	96 well	37.00	B2	15.00	11	1.4	
	8	13	10x	up/down 2x	96 well	25.00	B1	15.00	13	1.1	
	5	14	Simple	up/down 2x	12 well	25.00	B1	15.00	15	1.6	

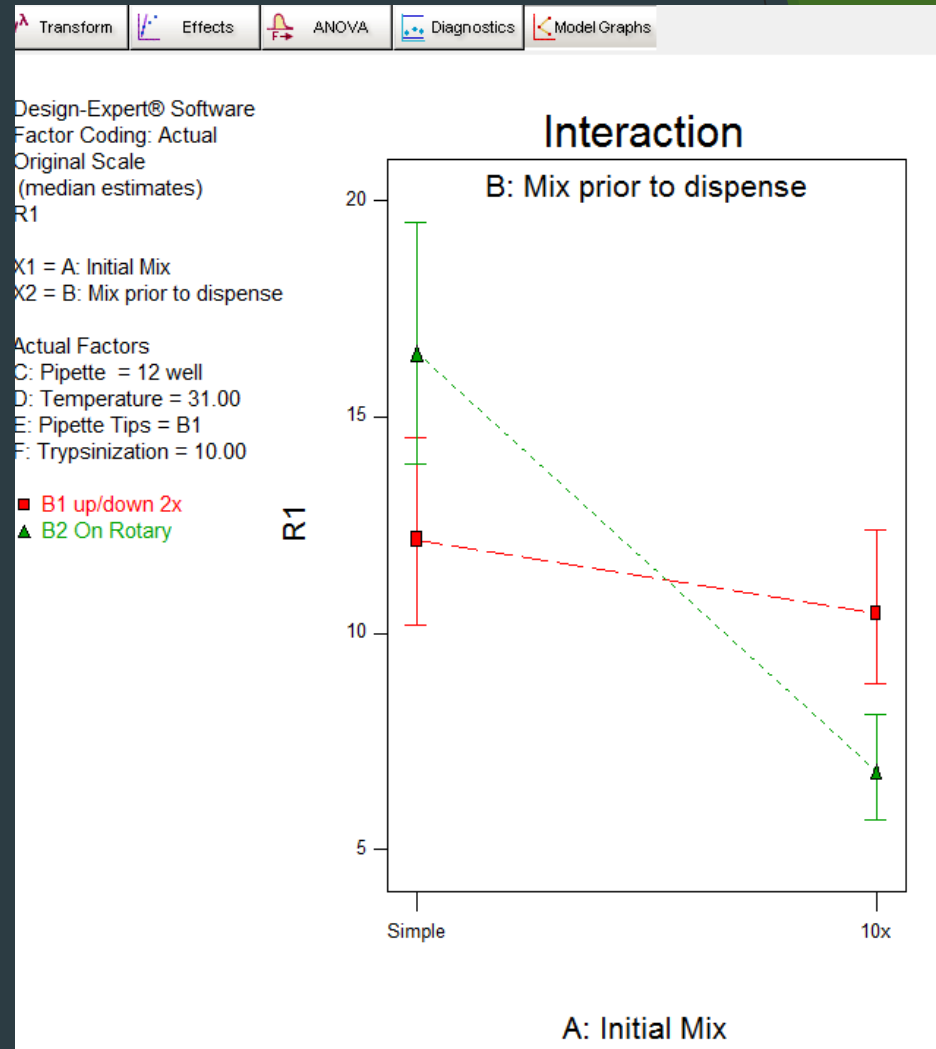
# R1 - Diagnostic Plot



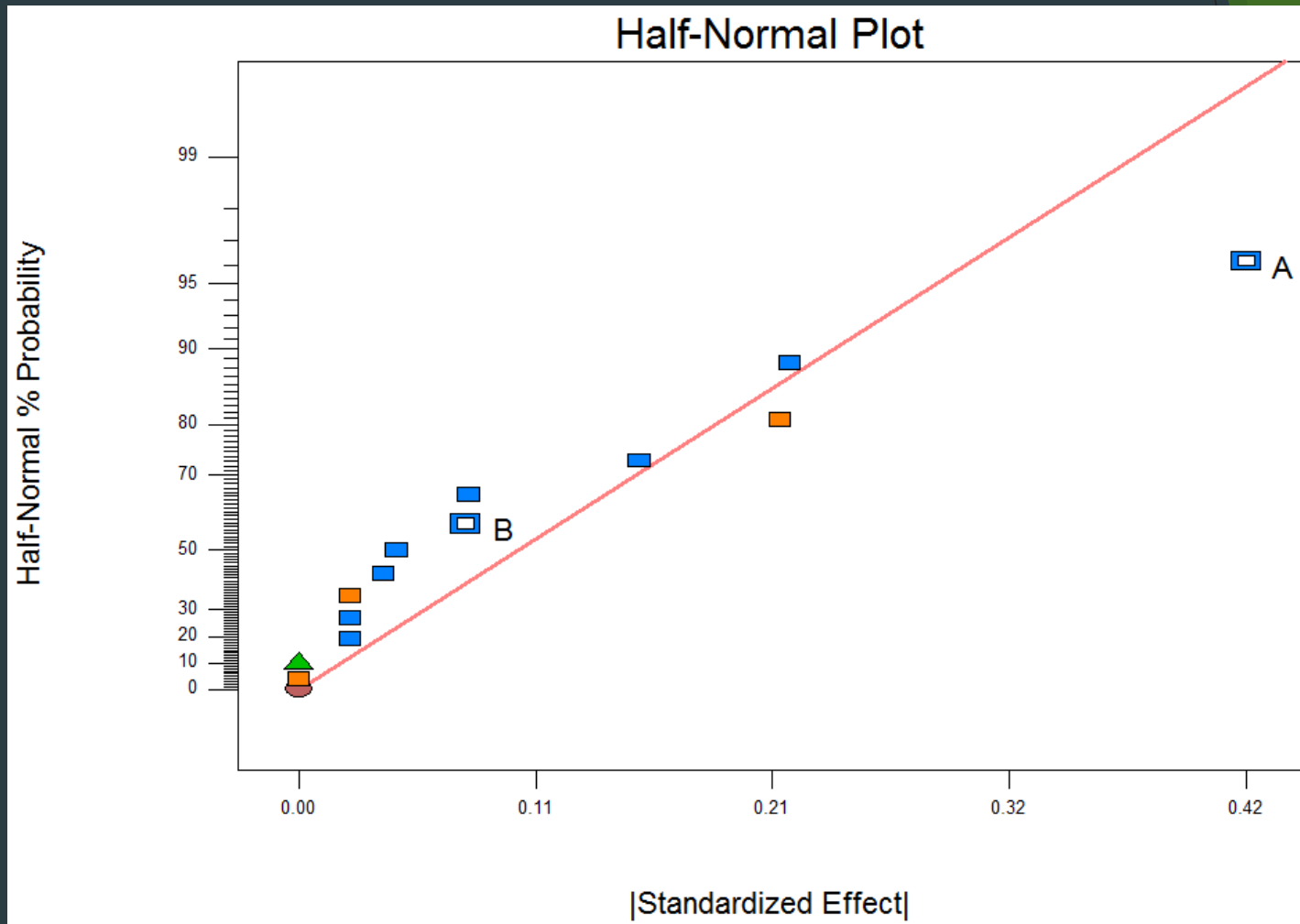
# R1 - Diagnostic Plot



# This Demonstrates the Importance of the Interaction



# Based on R2 (Ratio of 1<sup>st</sup> to Last Row)



# Now Went Back to the First Design

<b>Name</b>	<b>Units</b>	<b>Type</b>	<b>Low</b>	<b>High</b>
Passage #	passage	Factor	<P10	>P26
Seeding Density	cells per cm2	Factor	2000	4000
Tryp concentration	ml	Factor	2	4
Tryp Incub	Minutes	Factor	5	10
FBS lots	Lot	Factor	B1	B2
Day Feeding	times per week	Factor	1	2
R1	%CV*	Response		
R2	Bowl Ratio**	Response		

# Results

- ▶ Found pre-mixing (mixing prior to dispensing) important.
- ▶ Repeated the first screening and found passage number and trypsinization conditions were also important
- ▶ Did a  $2^3$  full factorial and limited the passage number to less than 20 (based on other available data - FACs studies to look at stability of the receptor expression).

# Design Selected

File Edit View Display Options Design Tools Help Tips

Combined  
Mixture  
Response Surface  
Factorial

2-Level Factorial  
Min-Run Res V  
**Min-Run Res IV**  
Irregular Fraction  
General Factorial  
Optimal  
Plackett-Burman  
Taguchi OA

## Minimum-Run Equireplicated Res IV Design

Design for 5 to 50 factors where each factor is varied over only 2 levels. Resolution IV design estimation of main effects. Two-factor interactions will be aliased with other two-factor and higher-order interactions. Good designs to reduce the number of runs if interactions are unlikely.

Factors:  (5 to 50)  Horizontal  Vertical

	Name	Units	Type	Low	High
A [Numeric]	Passage	P	Numeric	10	26
B [Numeric]	Seeding Densi	Cells/cm2	Numeric	2000	4000
C [Numeric]	Tryp concent	ml	Numeric	2	4
D [Numeric]	Tryp Incub	Minutes	Numeric	5	10
E [Numeric]	Feeding	Day	Numeric	1	2
F [Categoric]	BSA lot		Categoric	L1	L2

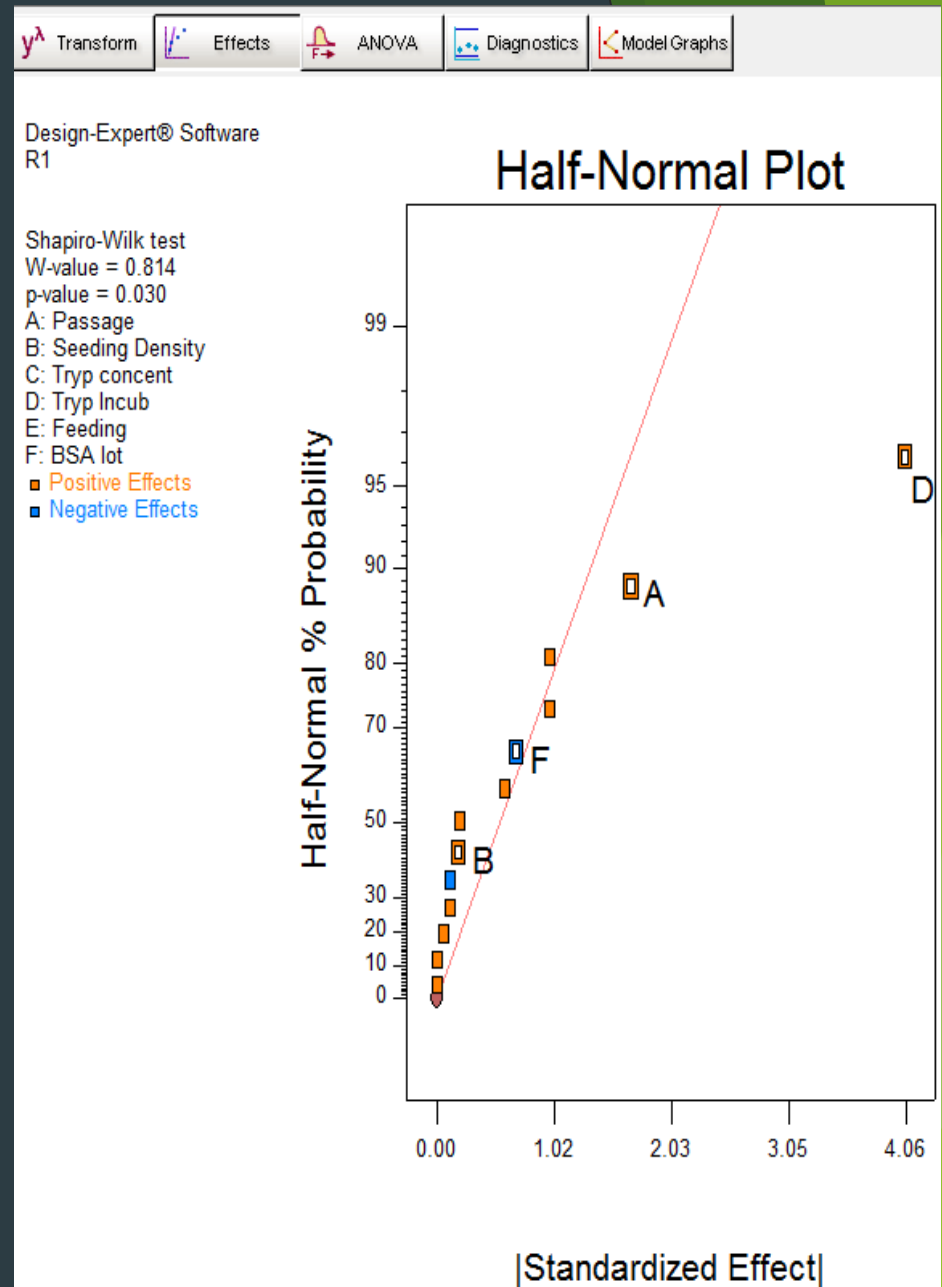


# Sorted by Passage Number

File Edit View Display Options Design Tools Help Tips										
Select	Std	Run	Factor 1 A:Passage P	Factor 2 B:Seeding D... Cells/cm2	Factor 3 C:Tryp conc... ml	Factor 4 D:Tryp Incub Minutes	Factor 5 E:Feeding Day	Factor 6 F:BSA lot	Response 1 R1 %CV	
	13	2	10.00	4000.00	2.00	5.00	1.00	L1	6	
	4	3	10.00	2000.00	4.00	5.00	2.00	L1	6	
	3	8	10.00	4000.00	4.00	10.00	1.00	L2	10	
	2	9	10.00	2000.00	2.00	10.00	1.00	L1	9	
	7	11	10.00	4000.00	2.00	5.00	2.00	L2	4	
	1	12	10.00	4000.00	2.00	10.00	2.00	L1	9	
	5	14	10.00	2000.00	2.00	5.00	1.00	L2	6	
	11	1	26.00	4000.00	2.00	10.00	1.00	L2	11	
	12	4	26.00	4000.00	4.00	10.00	2.00	L1	13	
	6	5	26.00	2000.00	4.00	10.00	2.00	L2	11	
	10	6	26.00	2000.00	2.00	5.00	2.00	L1	8	
	8	7	26.00	2000.00	4.00	5.00	1.00	L2	6	
	14	10	26.00	2000.00	4.00	10.00	1.00	L1	10	
	9	13	26.00	4000.00	4.00	5.00	2.00	L2	6	

# Diagnostic Plot

Indicates that the Passage Number and the Length of the Trypsin Incubation have an Effect



# Choose Model vs. Error Terms

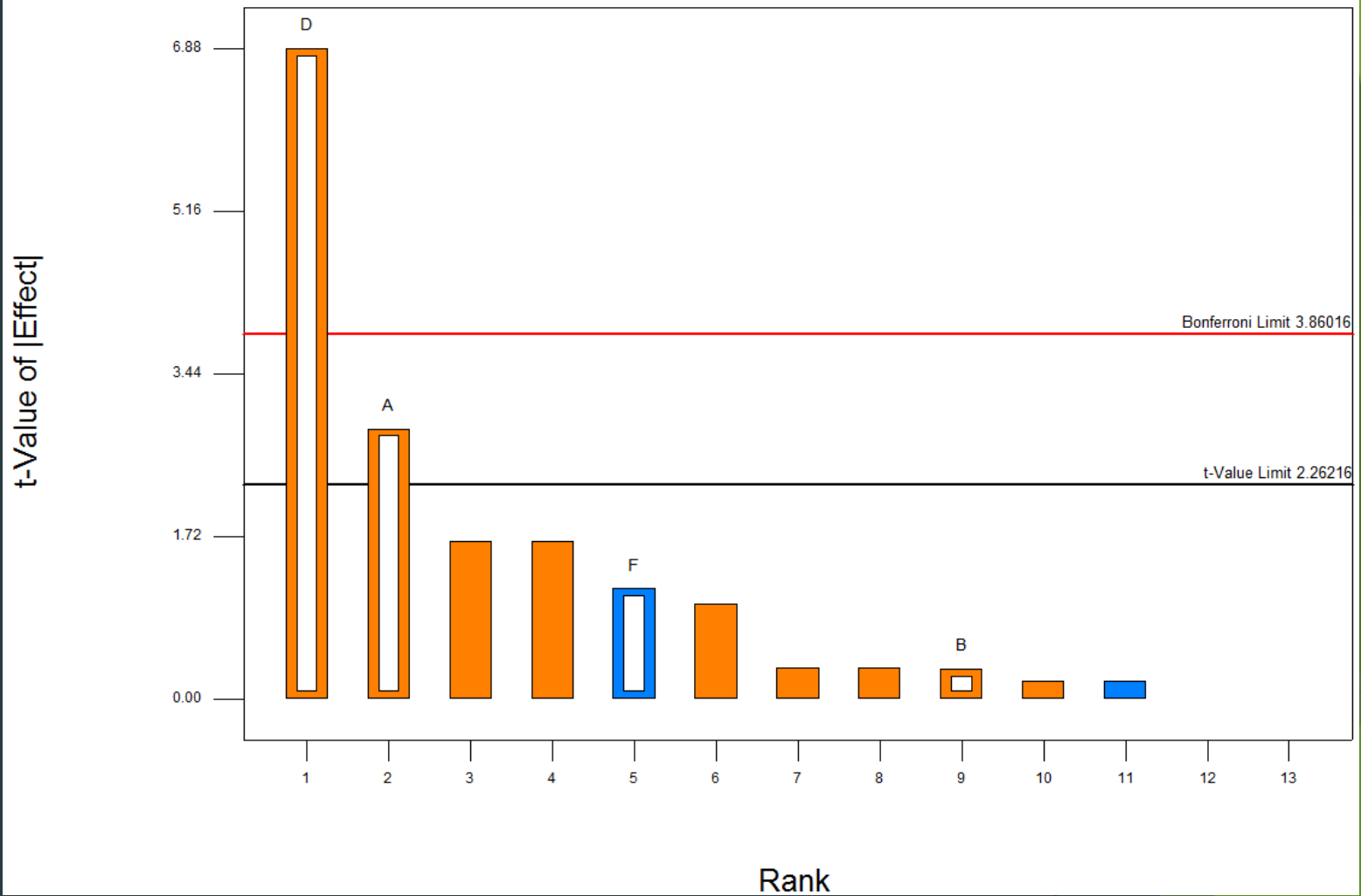
Notes for MyDesign

- Design (Actual)
  - Summary
  - Graph Columns
  - Evaluation
- Analysis
  - R1:R1
  - Optimization
    - Numerical
    - Graphical
  - Point Prediction
  - Confirmation

Selection:  Order:

	Term	Stdized Effects	Sum of Squares	% Contribution
	Intercept			
M	A-Passage	1.63	16.07	18.19
M	B-Seeding Density	0.20	1.93	2.18
e	C-Tryp concent	0.12	0.79	0.90
M	D-Tryp Incub	4.24	58.01	65.65
e	E-Feeding	0.065	0.20	0.22
M	F-BSA lot	-0.72	1.38	1.56
e	AB	0.000	0.000	0.000
e	AC	0.000	0.000	0.000
e	AD	0.65	2.11	2.39
e	AE	1.08	3.15	3.56
e	AF		Aliased	
e	BC		Aliased	
e	BD	1.08	4.41	4.99
e	BE	-0.13	0.19	0.21
e	BF		Aliased	
e	CD		Aliased	
e	CE		Aliased	
e	CF		Aliased	
e	DE		Aliased	
e	DF		Aliased	
e	EF		Aliased	
e	ABC		Aliased	
e	ABD	0.22	0.13	0.14

# Pareto Chart



# Take Home Messages

- ▶ This example was to indicate that the DOE can be used during optimization - not just final characterization or verification of assay performance.
- ▶ Sequential DOE studies is an excellent approach.
- ▶ Don't panic if the first design doesn't yield "results" - perhaps you didn't select the appropriate factors to study.
- ▶ Although a statistician is a real asset with modern software - you can still use DOE and have it really accelerate your development time.

# What to Do?

- ▶ This specific result should not be taken as a universal decision.
  - ▶ Specifically, not every cell line should be pre-mixed as described here.
  - ▶ Not every cell line will be sensitive to Trypsinization or have the same passage number restrictions.
- ▶ I would suggest instead, that you might be able to come up with a universal **screening design** - of factors into which specific levels for a given cell line could be inserted.
- ▶ Then, require that as part of development, each new proposed cell-line would be tested in this universal design.

# But.....DOE isn't the Panacea for all Component Assay Woes

It is a great tool to quickly differentiate critical parameters from those which have little impact.

It doesn't eliminate the need for focused scientific problem solving.

Always start with potential scientific root causes to performance problems

# But....What about the Following?

0.134	0.354	0.376	0.322	0.325	0.377	0.322	0.366	0.378	0.331	0.371	0.378
0.167	0.385	0.389	0.366	0.377	0.378	0.389	0.385	0.389	0.389	0.403	0.399
0.201	0.322	0.325	0.389	0.389	0.399	0.368	0.322	0.325	0.366	0.329	0.345
0.165	0.389	0.403	0.354	0.401	0.345	0.399	0.366	0.377	0.401	0.377	0.345
0.145	0.366	0.377	0.389	0.389	0.389	0.389	0.389	0.389	0.403	0.325	0.389
0.089	0.389	0.389	0.321	0.355	0.399	0.401	0.354	0.401	0.377	0.377	0.399
0.033	0.401	0.379	0.333	0.378	0.37	0.399	0.374	0.328	0.389	0.389	0.387
0.147	0.399	0.366	0.355	0.373	0.356	0.378	0.329	0.383	0.379	0.369	0.358

One Low Side





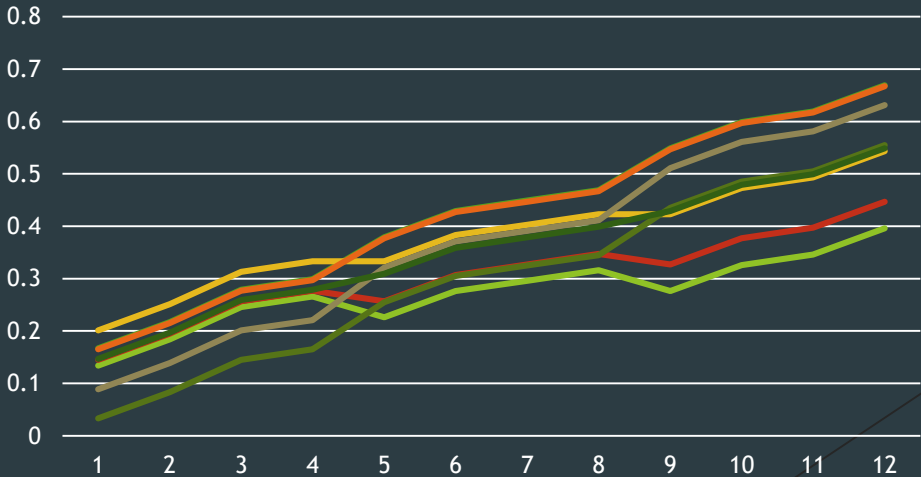
# Potential Causes

- ▶ Poorly calibrated pipette
- ▶ Insufficient Number of Cells
- ▶ Too concentrated of some biological component which is killing cells or inhibiting growth
- ▶ Reader Problem (off-set detector)

# Yet another case....

0.134	0.184	0.246	0.266	0.226	0.276	0.296	0.316	0.276	0.326	0.346	0.396
0.167	0.217	0.279	0.299	0.379	0.429	0.449	0.469	0.549	0.599	0.619	0.669
0.201	0.251	0.313	0.333	0.333	0.383	0.403	0.423	0.423	0.473	0.493	0.543
0.165	0.215	0.277	0.297	0.377	0.427	0.447	0.467	0.547	0.597	0.617	0.667
0.145	0.195	0.257	0.277	0.257	0.307	0.327	0.347	0.327	0.377	0.397	0.447
0.089	0.139	0.201	0.221	0.321	0.371	0.391	0.411	0.511	0.561	0.581	0.631
0.033	0.083	0.145	0.165	0.255	0.305	0.325	0.345	0.435	0.485	0.505	0.555
0.147	0.197	0.259	0.279	0.309	0.359	0.379	0.399	0.429	0.479	0.499	0.549

Gradual Increase Across the Plate



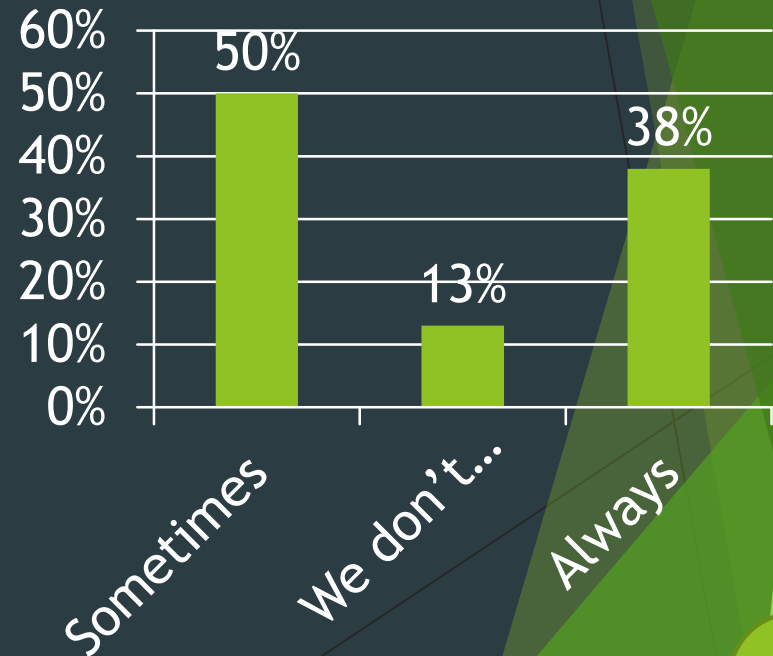
# Possible Causes

- ▶ Settling of cells during initial pipetting
- ▶ Fragile cells breaking during mixing during plating
- ▶ Increased or Decreased concentration of a critical component because of dilution scheme
- ▶ Time differences due to manipulation of the cells

# Your Components

We have talked a lot about cells. Do you optimize well-to-well characteristics of your plates?

1. Sometimes
2. We don't usually need to because we get our cells from a potency group which has optimized our cells
3. Always

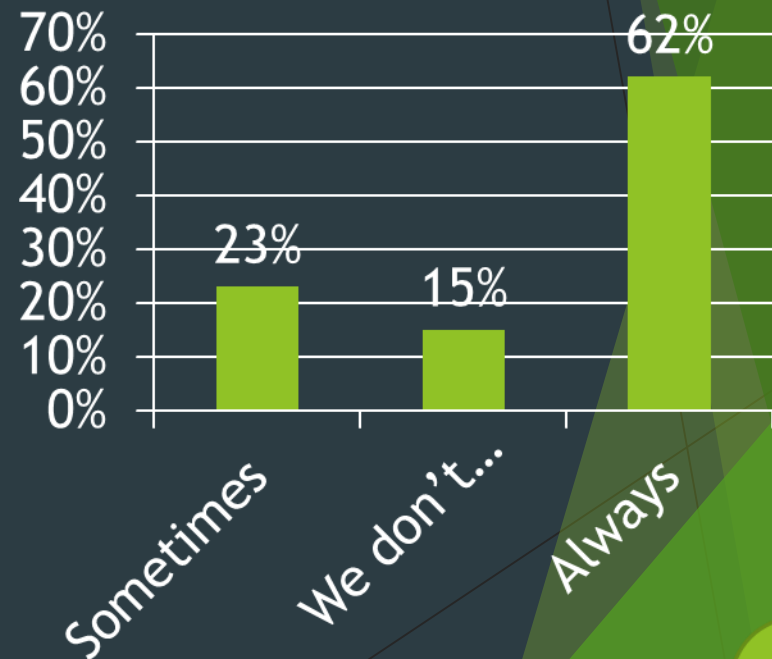


# Your Components

EU Response

We have talked a lot about cells. Do you optimize well-to-well characteristics of your plates?

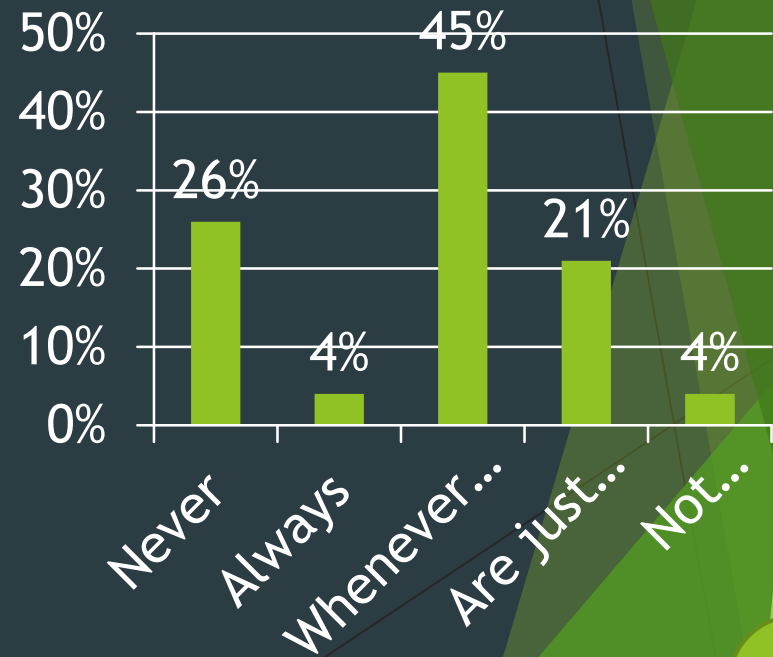
1. Sometimes
2. We don't usually need to because we get our cells from a potency group which has optimized our cells
3. Always



# Ready-to-Use

Do you use Ready to Use cells?

1. Never
2. Always
3. Whenever possible
4. Are just implementing
5. Not applicable to our products

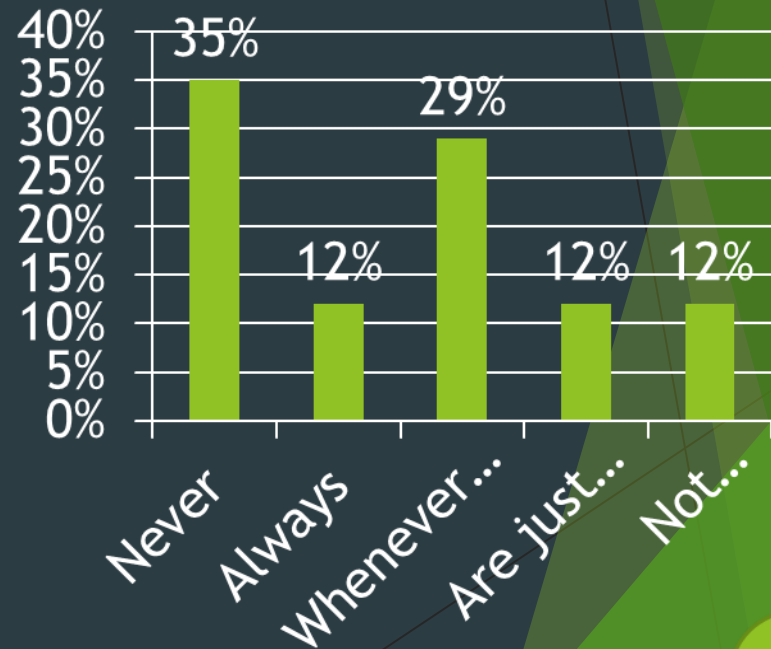


# Ready-to-Use

## EU Response

Do you use Ready to Use cells?

1. Never
2. Always
3. Whenever possible
4. Are just implementing
5. Not applicable to our products

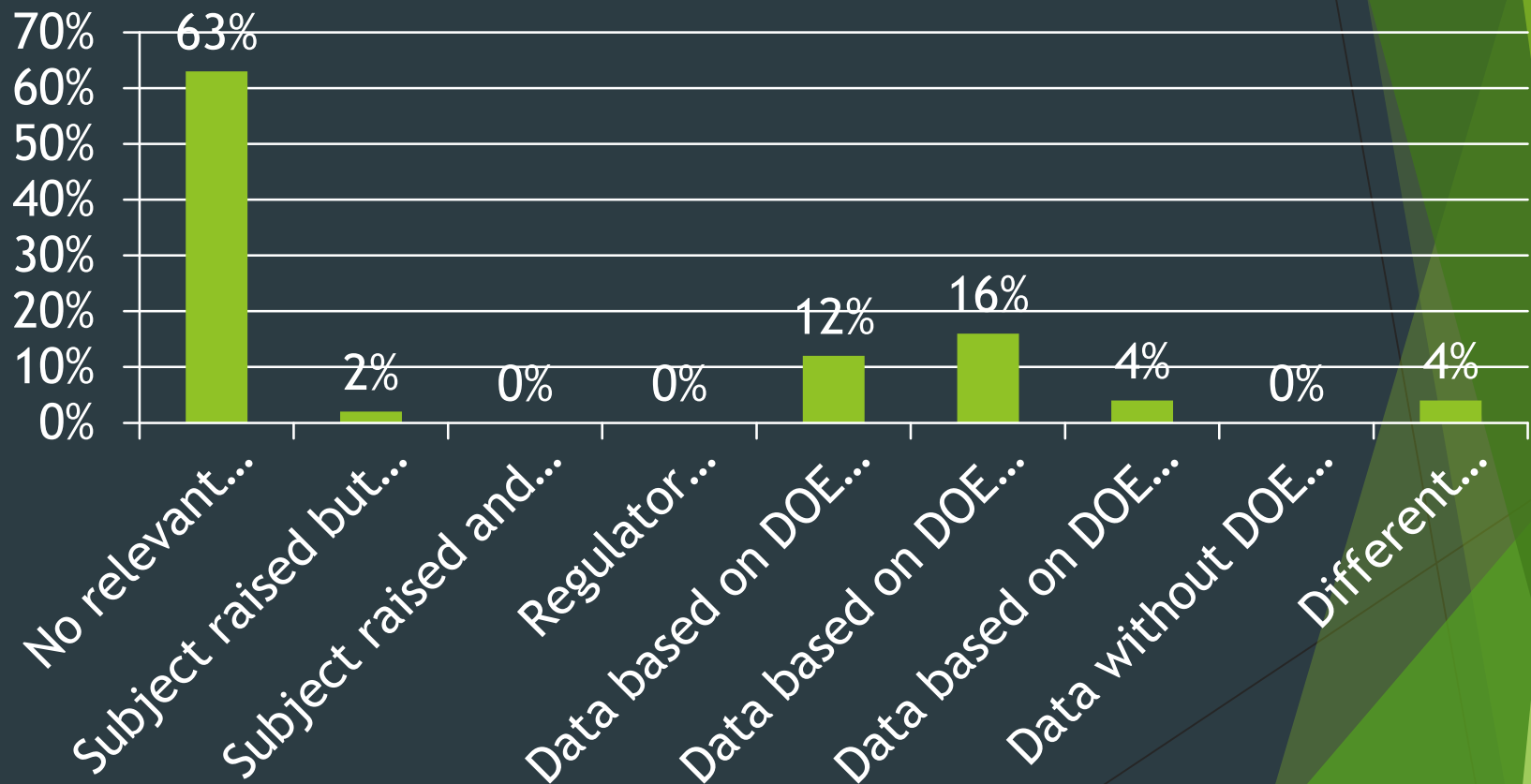


# Have you had any discussion on DOE with regulators?

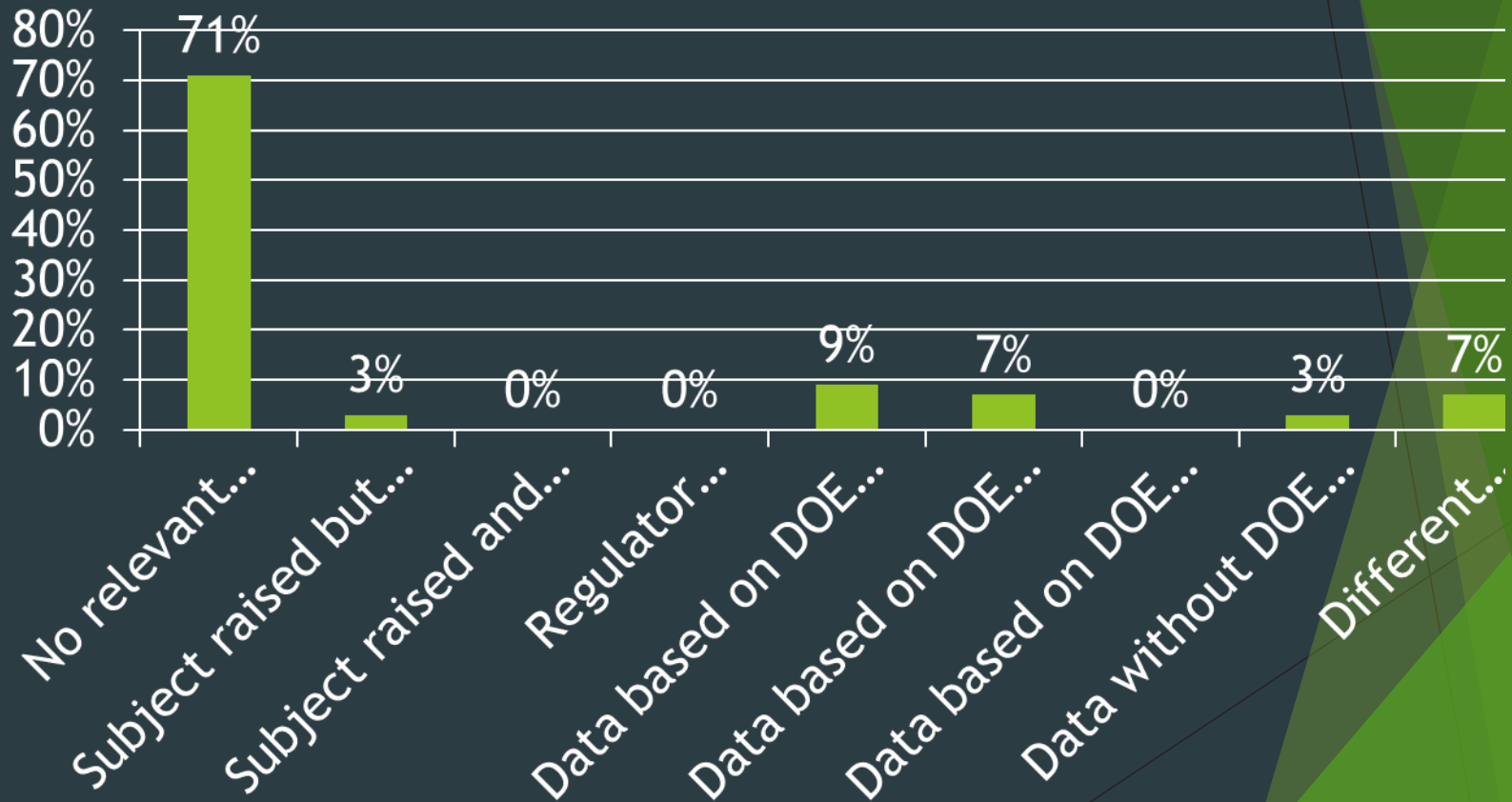
1. No relevant discussion with regulator
2. Subject raised but no comment from regulator
3. Subject raised and regulator suggested use
4. Regulator spontaneously suggested use
5. Data based on DOE submitted - no comment
6. Data based on DOE submitted - favourable response
7. Data based on DOE submitted - modifications suggested
8. Data without DOE submitted - regulator required use
9. Different responses in different cases



# Discussion with Regulators



# Discussion with Regulators (EU Response)



# Appendix: DOE Basics

These are for your convenience and were not covered in the talk.

# DOE Designs

- ▶ If you only have a few conditions (such as with stability) - typically do **full-factorial designs**
- ▶ If you have many conditions and you are interested in finding which (if any) are important, you will do partial factorial (screening) designs
  - ▶ E.g. **Plackett-Burman, Fractional Factorials**

# DOE - Basics

- ▶ **Factors** = The assay conditions or reagents that you vary (e.g.: incubation temperature, dilution, incubation time, etc. Usually assign these letters
- ▶ **Level** = The condition of the factor which you test (e.g.: 25 minutes vs. 35 minutes, 1:1000 vs. 1:2000 dilution, etc.) In the following example:

**High level = Plus (+) (or can be capital letters)**

**Low Level = Minus (-) (or can be small letters)**

# Example # 1

- ▶ Few Conditions: Therefore = factorial design → look at three factors → at two levels:
- ▶  $2^3$  → This is three factors
- ▶ This tells you there are two levels.

An example would be :



Factor	-	+
Time	2h	4h
Temperature	20° C	37° C
pH	6.5	8

# What is Factorial Design?

- ▶ Set of experiments so that more than one variable can be tested at the same time.
- ▶ This is done by running all the possible combinations-of each factor at each level.
- ▶ Therefore  $2^2 = 2*2$  experiments: 4 experiments
- ▶ And  $2^3 = 2*2*2$  experiments: 8

Note that all of these designs are balanced. This is a key aspect of DOE

$2^2$

$X_1$	$X_2$
-	-
+	-
-	+
+	+

4 combinations that will be run

$2^3$

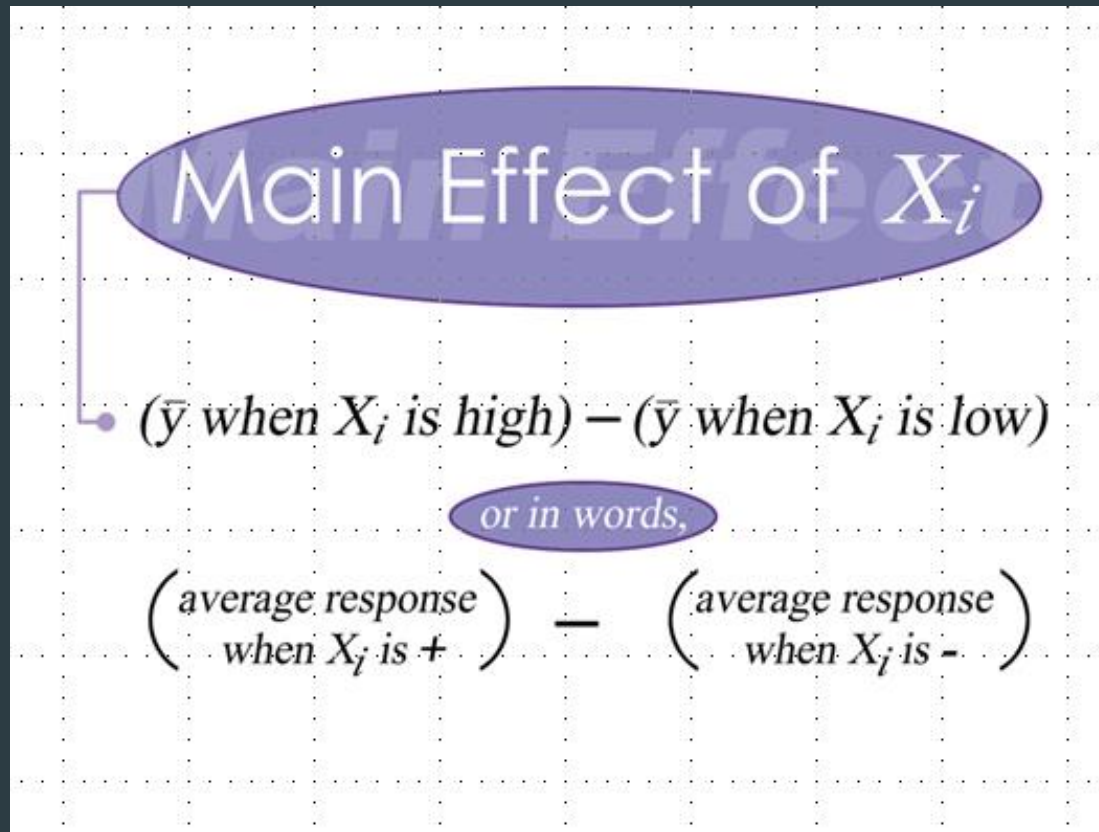
$X_1$	$X_2$	$X_3$
-	-	-
+	-	-
-	+	-
+	+	-
-	-	+
+	-	+
-	+	+
+	+	+

8 combinations that will be run

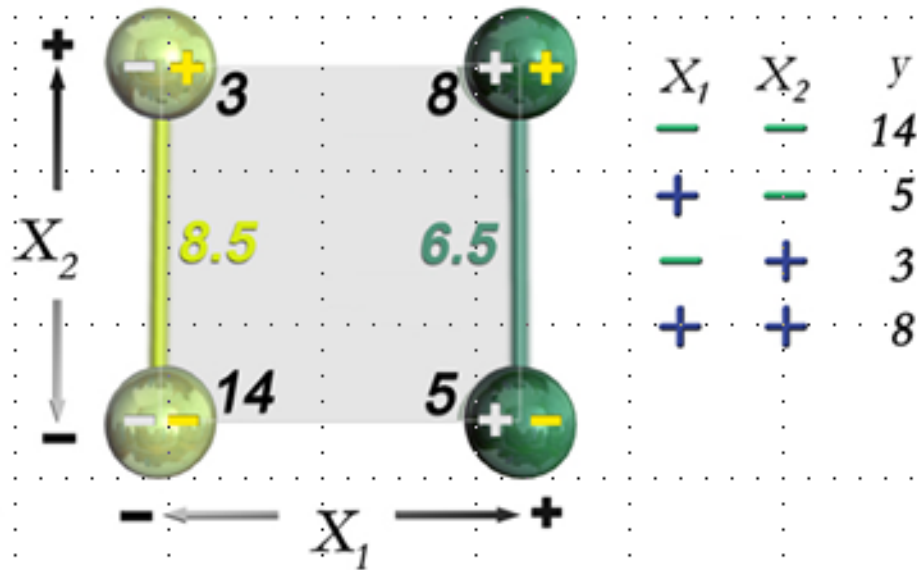


This type of design allows us to determine the effect of changing each variable (aka: the main effect):

NB: Y is the measured response



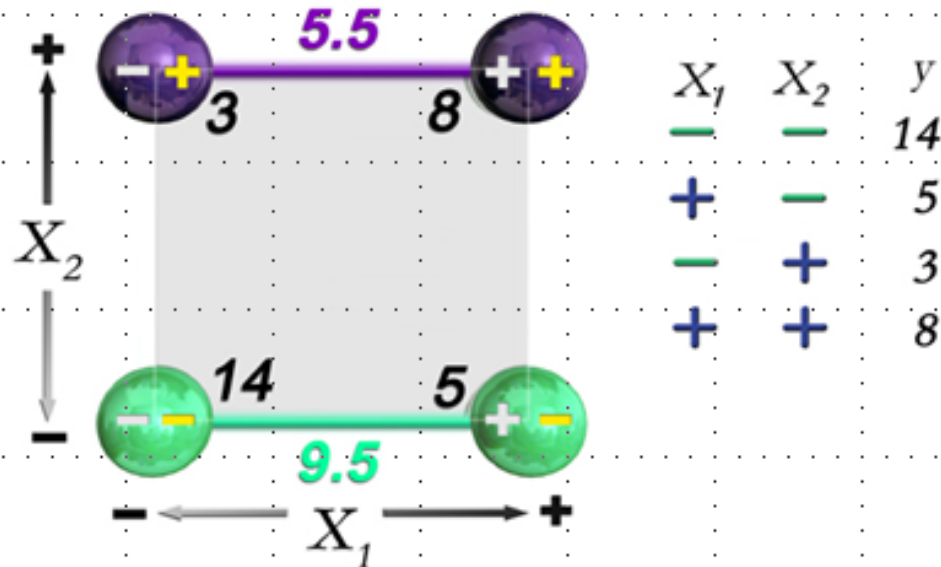
## Main Effect of $X_1$



$$E_1 = \left( \frac{8 + 5}{2} \right) - \left( \frac{3 + 14}{2} \right) = -2$$

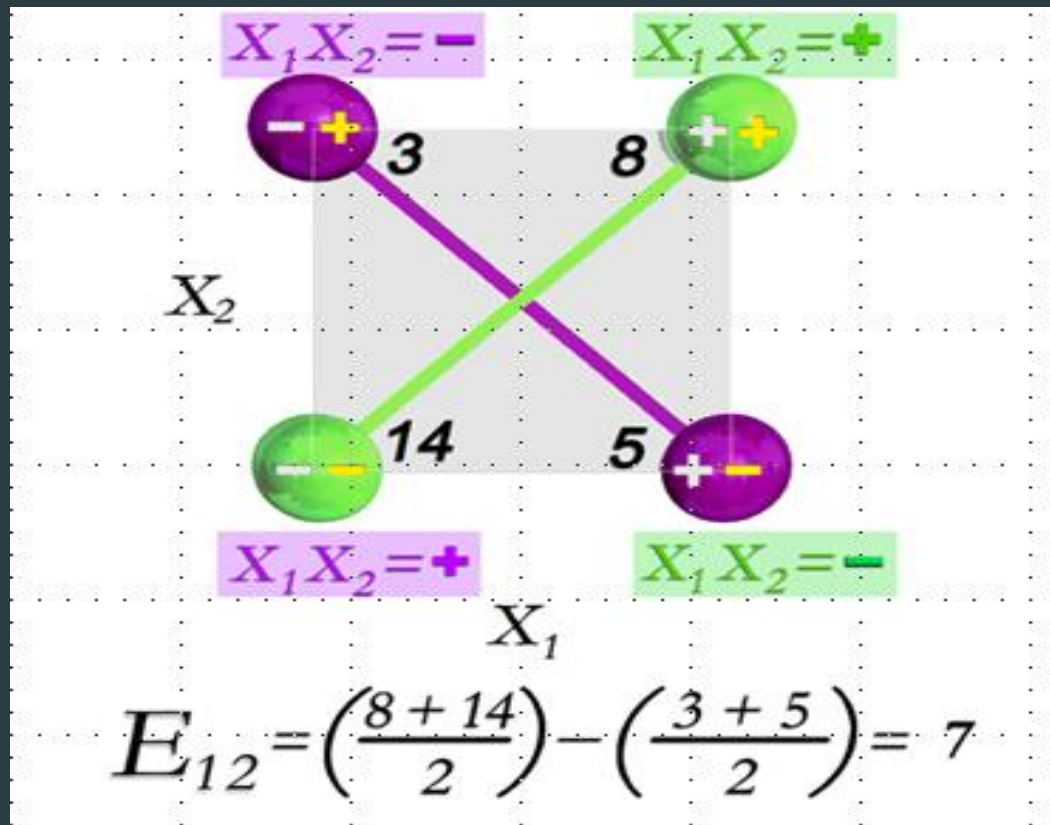
NB: E is a measure of the effect

## Main Effect of $X_2$



$$E_2 = \left( \frac{3 + 8}{2} \right) - \left( \frac{14 + 5}{2} \right) = -4$$

# The $X_1X_2$ Interaction



# What is the Problem?

- ▶ Early in development - we have many variables.
- ▶ A full-factorial design for many variables soon becomes too big.
- ▶ An example: Cell growth:

Media Type	Location in incubator
%FBS	Initial thawing temperature
Seeding Density	dispersion technique
Feeding Schedule	Maximum # of passages
Method of removing cells	Culture Time

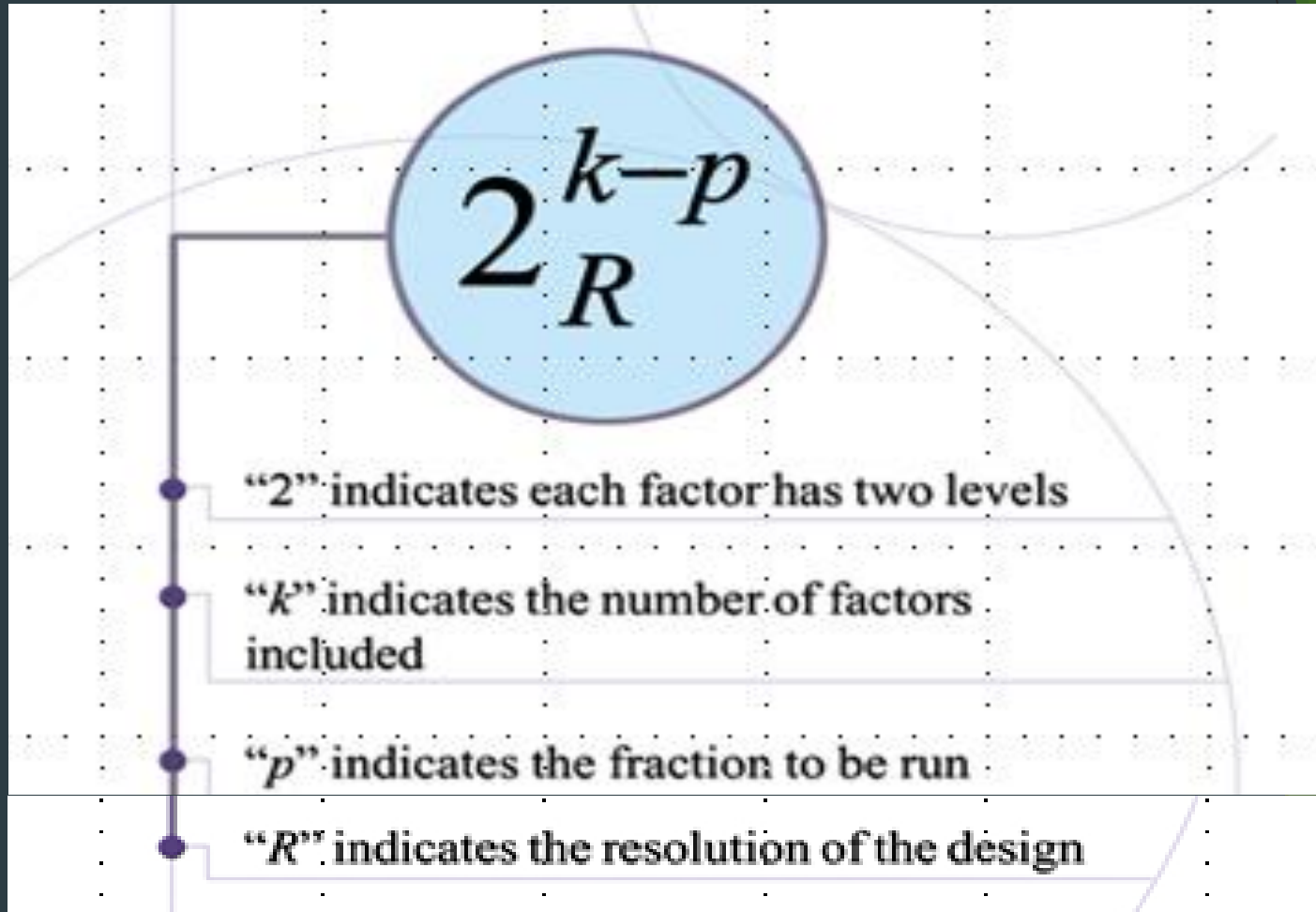
# Problem (continued)



- ▶ This simple example leads to:  $2^{10} = 1024$  experiments!!!!
- ▶ This is where the fractional factorial comes in.

A Fractional Factorial Design runs a subset of the full factorial runs. If chosen correctly, we can still estimate the main effects but may lose the higher order interactions.

# Fractional Factorial Design



# Laureen's slide Your Designs(#11)

What kind of designs do you use?

1. Full Factorial only
2. Specific screening design only (such as a placket-burman)
3. Fractional Factorial
4. A mixture of the above
5. Too early in our use of DOE to be able to answer this question



# Future use of DOE in your lab

1. No plans to start
2. Plan to start
3. Plan to discontinue
4. Continue with limited use
5. Plan to expand use

# Laureen's slide Your Designs(#9)

How do you design your DOEs?

1. We have a statistician (either employed or consultant)
2. Use a software and design my own.
3. Design my own without software

# Laureen's slide Your Designs(#11)

What kind of designs do you use?

1. Full Factorial only
2. Specific screening design only (such as a placket-burman)
3. Fractional Factorial
4. A mixture of the above
5. Too early in our use of DOE to be able to answer this question