

Analysis of Test Results

Can We Minimize The Risk To The Geoprofessional Business?

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Introduction

About me

- Studied Engineering, Mathematics and Computer Science
- Did Graduate work in Entrepreneurship and Innovation
- With Spectra QEST since 1997
- Director and owner from 1998 to 2018
- Currently Head of Global Sales

Acknowledgment

- Krzysztof Kot presented a 'Data Discovery' Session in SQ User Conference
- Presented the importance of well-structured data and benefits thereof
- One idea stood out for me; so I developed it further



Well-Structured Data

What is it?

- Consider a sieve test
- Database sieve masses in order (Sieve_1, Sieve_2, etc. in stack)...
- ...or explicitly (for example, 'Sieve_30')
- Latter harder, but more powerful

Importance:

- Ability to review more data
- More data enables us to see trends

| Sieve Size | Non Cumulative Mass (g) | Max Mass (g) | Passing (%) |
|------------------------|-------------------------|--------------|-------------|
| Tray Mass | 163.0 | | |
| Tray + Wet Mass | 5736.0 | | |
| Wet Mass | 5573.0 | | |
| Tray + Dry Mass | 5328.0 | | |
| Dry Mass | 5165.0 | | |
| After Wash Mass | 5133.0 | | |
| 3/8in | 0.0 | - | 100 |
| No. 4 | 120.0 | - | 98 |
| No. 8 | 625.0 | - | 86 |
| No. 16 | 789.0 | - | 70 |
| No. 30 | 1259.0 | - | 46 |
| No. 50 | 1107.0 | -* | 24 |
| No. 100 | 844.0 | - | 8 |
| No. 200 | 385.0 | - | 0.7 |
| Pan | 10.0 | - | |
| Finer 75µm | 32.0 | | 0.6 |

| Sieve Size (mm) | Non Cumulative Mass (g) | Max Mass (g) | % Pass (total) |
|--------------------------|-------------------------|--------------|----------------|
| Tare Mass | | | |
| Tare+Vet | | | |
| Wet Mass | 3157.0 | | |
| Dry Mass | 2999.0 | | |
| 1in | 30.0 | 1800 | 99 |
| 3/2in | 84.2 | 1400 | 96 |
| 1/2in | 94.0 | 890 | 93 |
| 3/8in | 182.4 | 670 | 87 |
| No. 4 | 425.0 | 330* | 73 |
| Mass Before Split | 2183.4 | | |
| Split Mass | 408.0 | | |
| No. 10 | 69.0 | - | 60 |
| No. 16 | 70.0 | - | 48 |
| No. 30 | 93.5 | - | 31 |
| No. 40 | 83.4 | - | 19 |
| No. 50 | 94.4 | - | 14 |
| No. 100 | 23.4 | - | 10 |
| No. 200 | 35.0 | - | 4 |
| Pan | 22.7 | - | |

Risks are Real

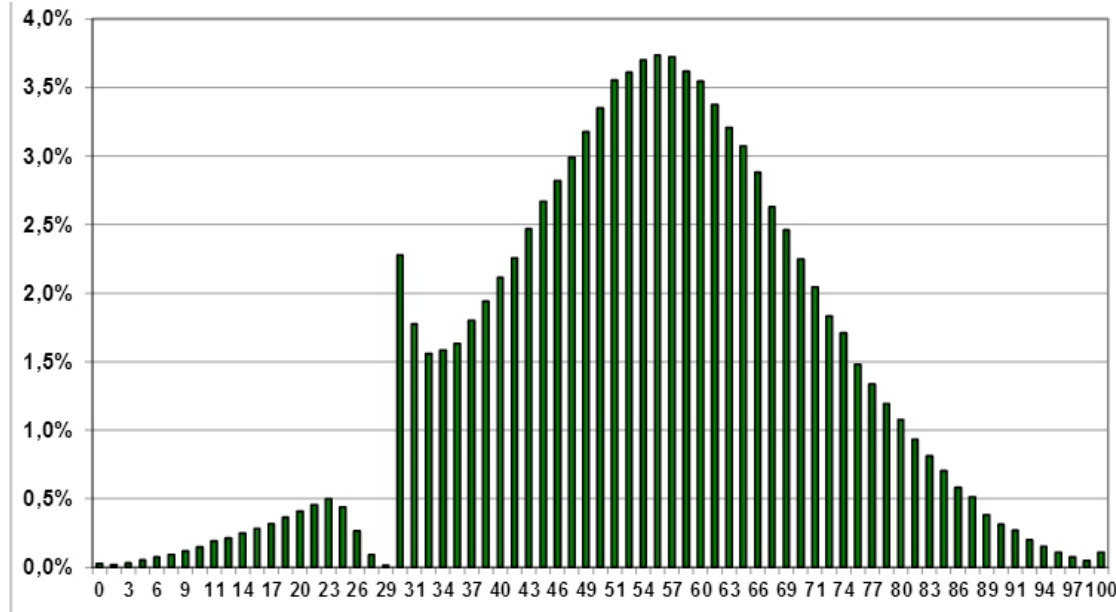
- And very scary!
- NY Company accused of falsifying results
 - Concrete testing, among other things
 - Management claimed they had no idea
 - President got 21 years in jail
- NC Company accused of false asphalt testing
 - 6 technicians involved
 - Company fined \$2.25M



The Idea

- Assume that management was not involved in deliberate data manipulation
 - Could they have done something to uncover the issue?
- **We believe there is – even though it is not a trivial matter!**
- Firstly, recognise that there is a human factor here...
 - ... who may either feel appropriate to pass 'close enough' results, or
 - ... who is simply not doing the 'right thing'
- Secondly, recognise there is interesting information in your testing database
 - Provided you have a well-structured database
 - You should be able to identify trends from existing data
- After all, **Big Data Means Big Opportunities!**

The Inspiration: HS Exam (Poland 2013)



Minimum score to pass: 30%

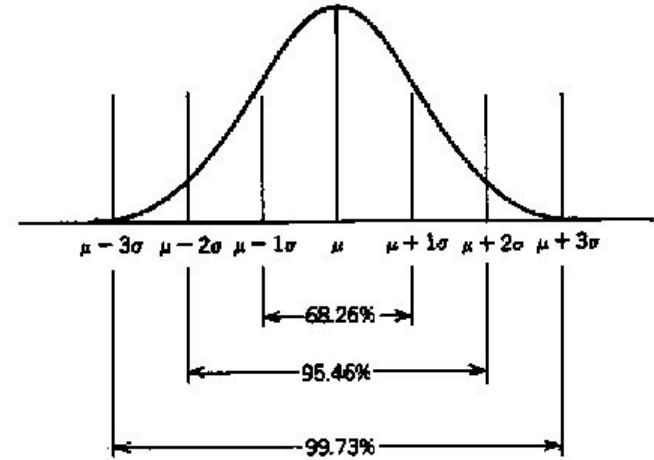
Correlation?

- We have human testers making essentially a similar determination
- Could they be displaying similar behaviour?
- Can we check?



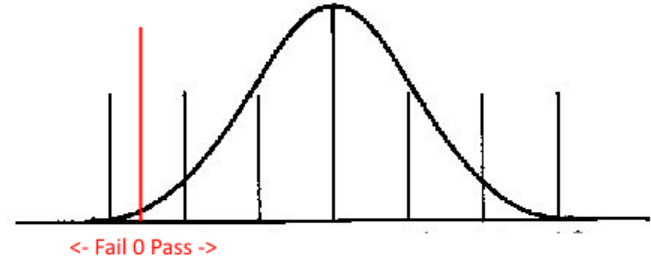
The Theory of Setting Specifications

- Consider high volume tests
 - Concrete strength
 - Field compaction percentage
- Expect normal distribution of results
 - Set μ such that your pass/fail point is somewhere at $\mu - k\sigma$
 - k is somewhere between 2 and 3 depending on your appetite for failures
- That's the theory
 - The practice relies on humans making decisions
 - But also, can the limit be set too aggressively?

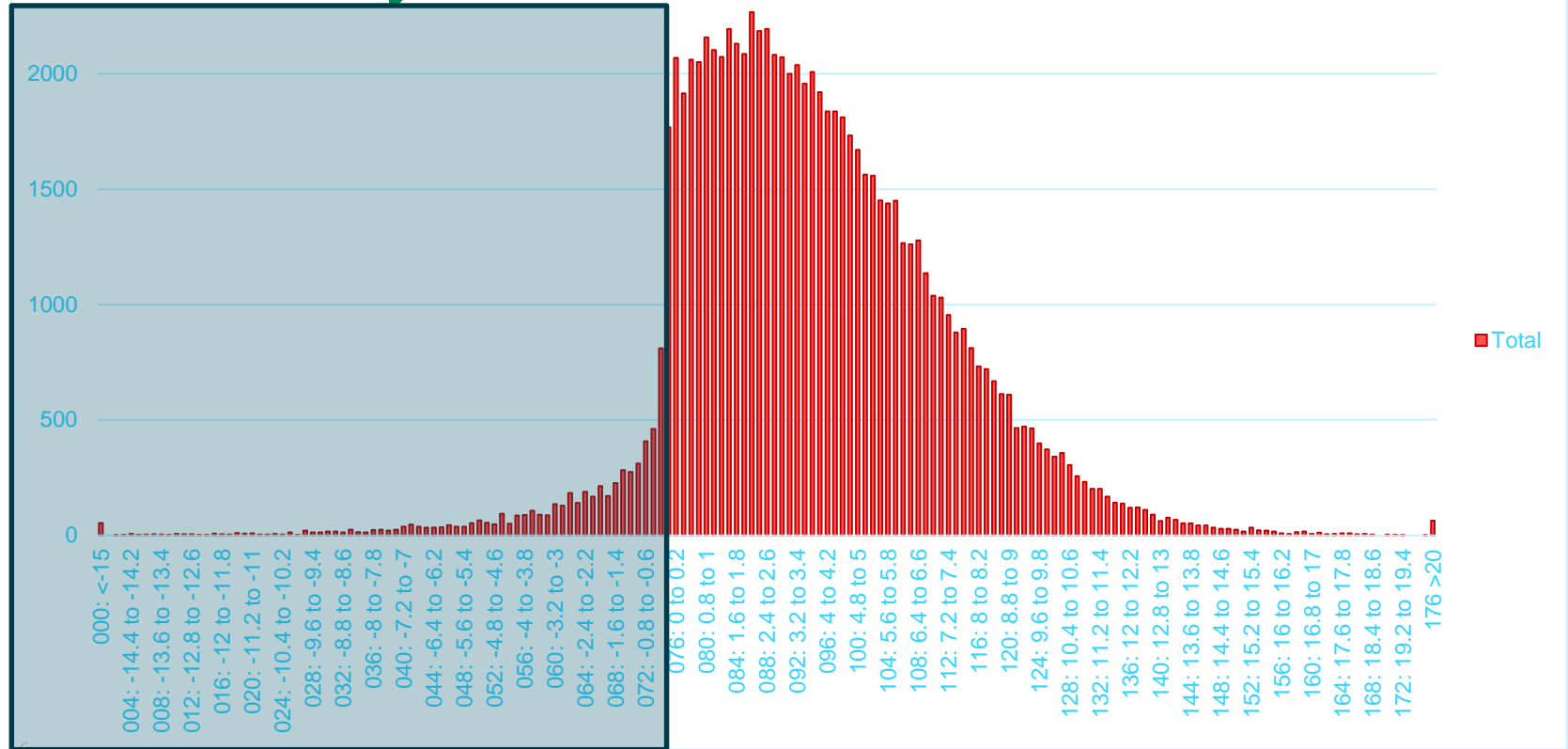


Normalising Results

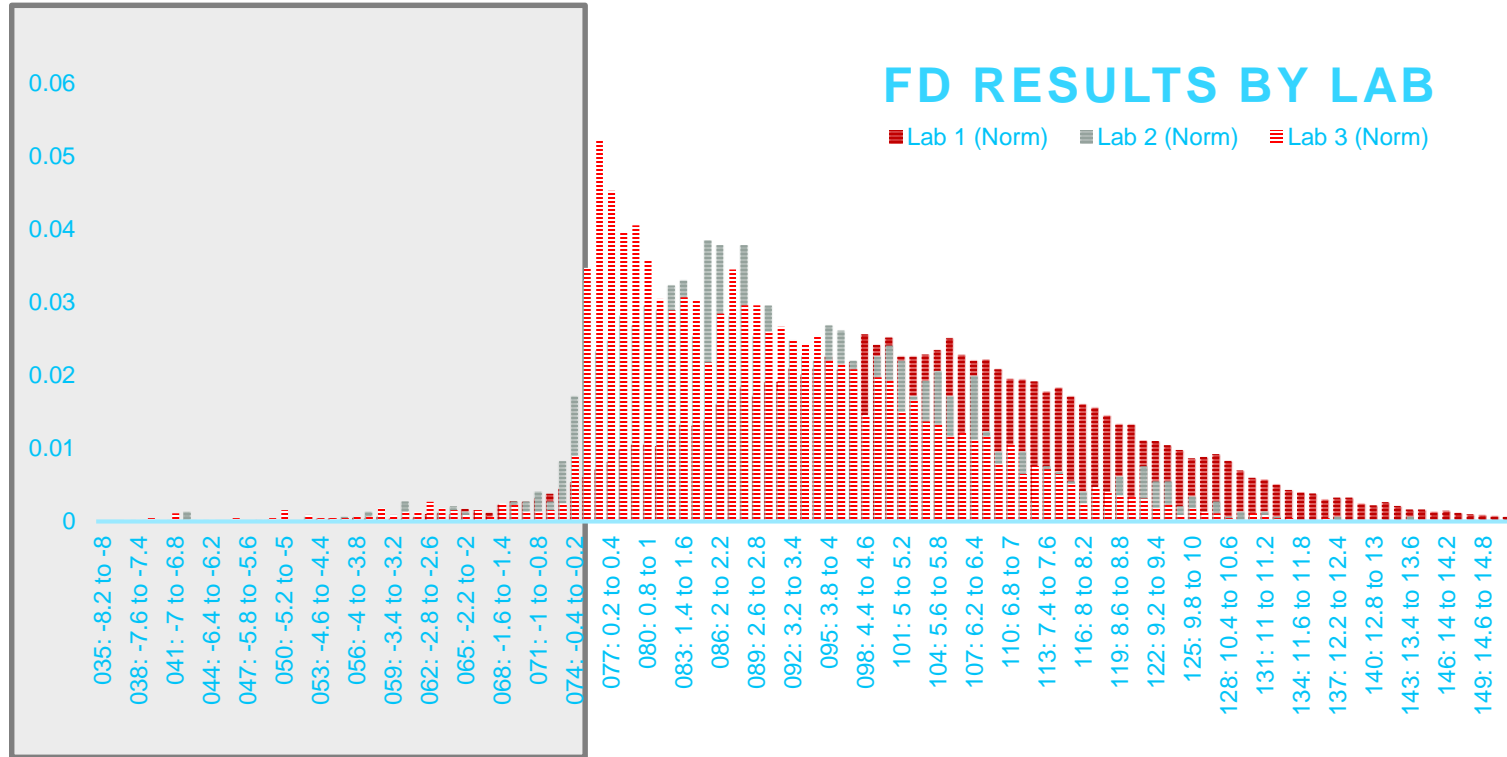
- Limits differ; How do we compare results?
- We can normalise results: Move limit to 0
- For concrete tests
 - Limit = F'_c
 - Consider: $(\text{Result} - \text{Limit}) / \text{Limit}$
- For Field Density tests
 - Limit = Some Ratio of (Adjusted) Max Dry Density
 - Consider: Result - Limit; (i.e. Compaction % - Required %)
- Let's take a look
 - Anonymised real data



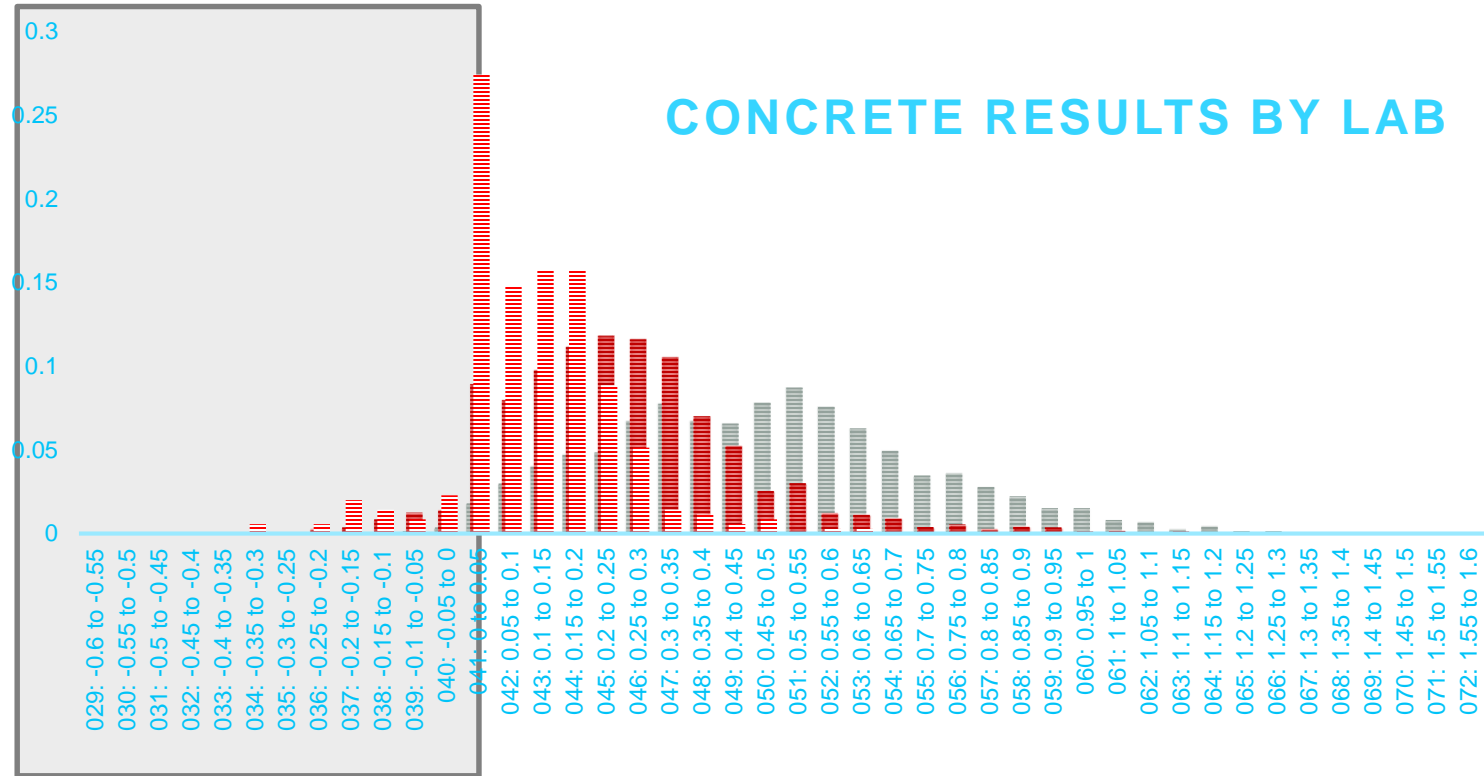
Field Density



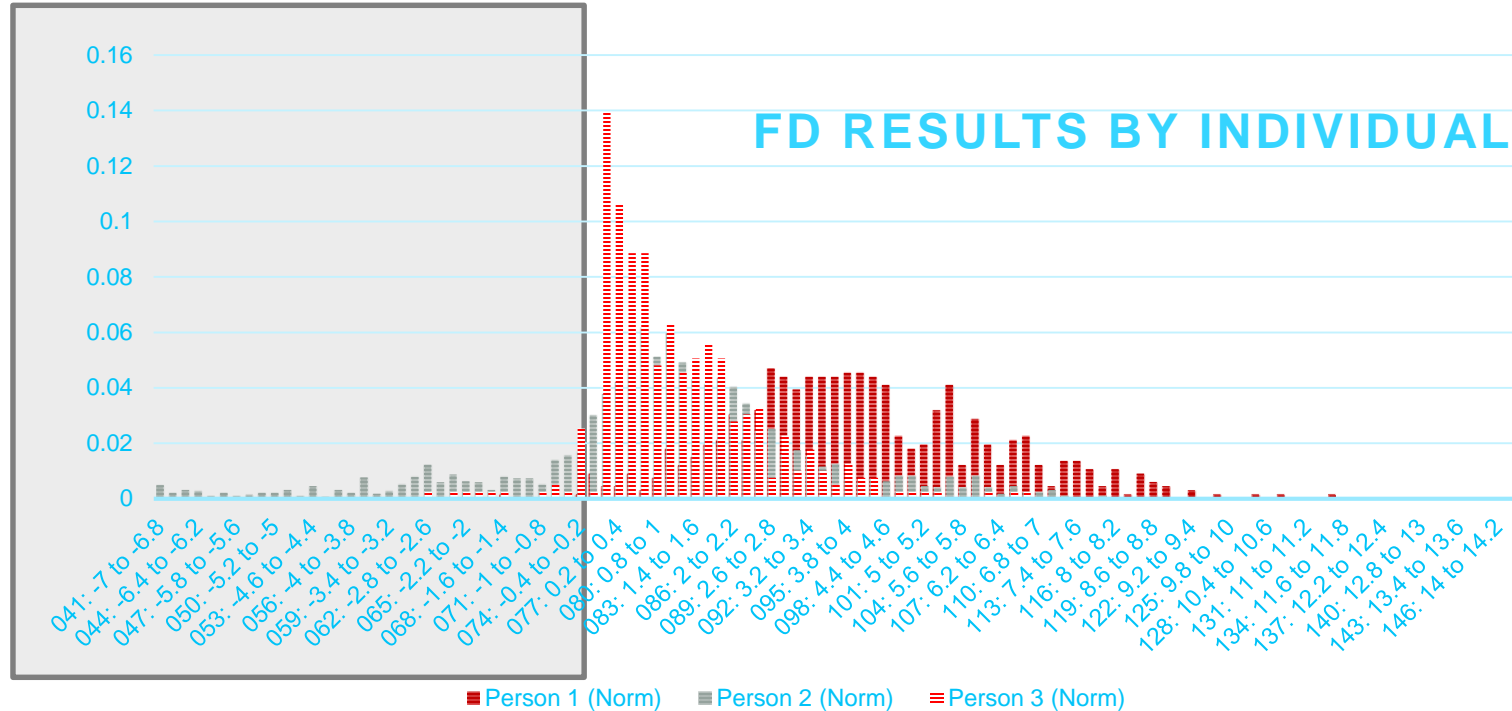
FD Results – By Laboratory



Concrete Results – By Lab



FD Results – By Individual



Summary

- Lots of Variability by Location and Individual
- Desired Result Distribution is Possible
- Interesting Observation: Most Problems Occur When Mean of Distribution is Closer to Pass/Fail – Indicates a Process Problem.

Prerequisites

- Well-structured data
- Enough test data
 - But perhaps not as many as you think; a few thousands are enough
- Concrete testing example
 - Test 50 cylinders/day → 250 in a week → More than 600 in a month
 - In three to four months we should have enough data for overall evaluation
 - In six months to a year, enough to start troubleshooting
- Field density example
 - Test 50 shots/day → 250 in a week → More than 1,000 in a month
 - In four months we should be able to evaluate overall
 - In less than a year, we should be able to troubleshoot

Test The Idea

- Try this idea with your data
- Can you find issues?
- Can you systematise this process to make it easy to run?



Recommendations

- **If you find the idea works with your data...**
- Identify centres of excellence
- Transfer knowledge and processes
- Identify locations/individuals for further training
- Use quality as a competitive advantage in the bid process
 - Quantitative proof of result validity: Dedication to quality
 - Can your competitors do that?
 - Will the market start requesting it?

Questions?



Thank you!

