

# Trend Detection in Car Manufacturing Processes

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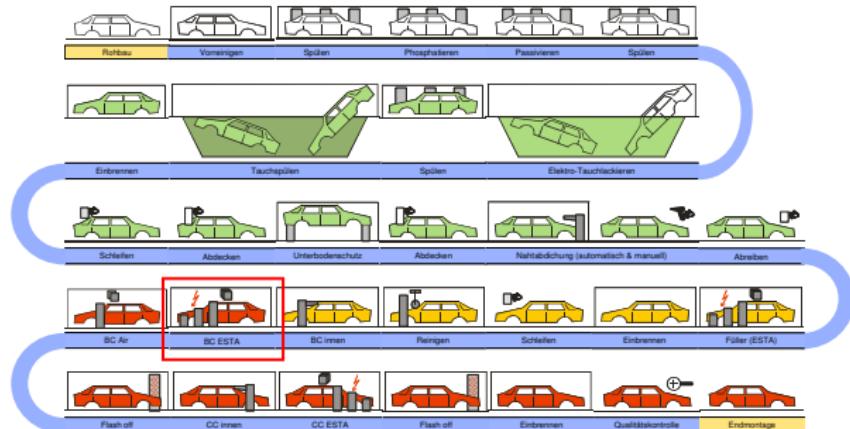
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# Car Production Today

- outsourcing, different suppliers involved
- painting is important cost factor
- cost(painting)>cost(body)



Trend Detection



## Problem

- early detection of trend needed
- construction of statistical tests
- comparison of test procedures



Source: BASF



# Outline

1. Motivation ✓
2. Dataset
3. Tests
4. Simulation

Trend Detection

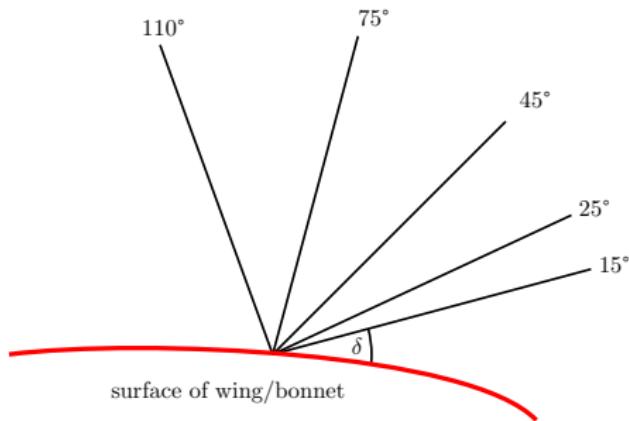


## Dataset

- 7100 painted cars, 20 different colors
- luminance, a- and b- color axis from five angles for bonnet and wing
- explanatory variables: color lot, temperature, measuring person
- 42% silver (luminance most important component)
- customers most sensitive to changes of metallic colors



## Used Angles



$\delta$	$L^*$	$a^*$	$b^*$
15°	98.94	1.58	4.43
25°	81.35	1.17	4.20
45°	50.98	0.73	2.58
75°	32.30	0.48	1.29
110°	26.11	0.21	0.87

Figure 1: Used angles and typical Lab-values for 'silver'



## The Lab Colorspace

- color model proposed by International Commission on Illumination
- each color defined by a tuple of luminance and the positions between red-green ( $a$ -axis) and blue-green ( $b$ -axis)

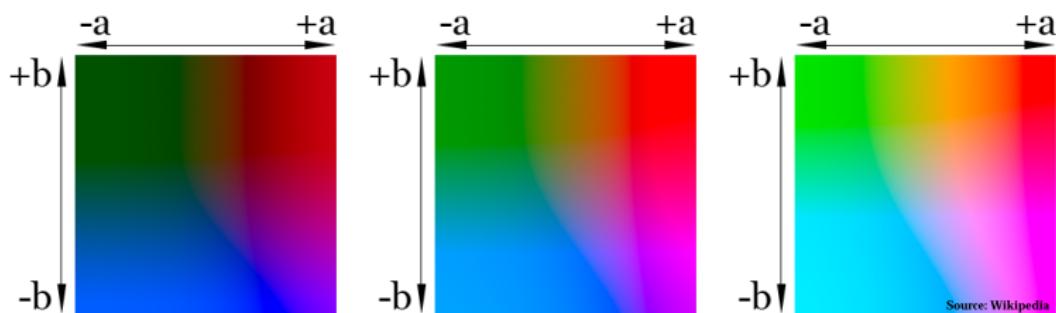


Figure 2: Color tiles for increasing values of luminance, source: Wikipedia



## Data Example

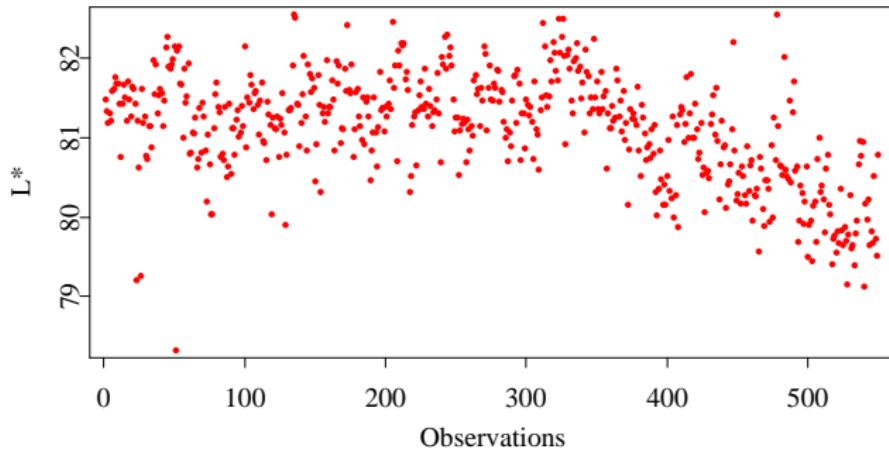


Figure 3: Sequence of L\*-values for 550 'silver' observations



## Proposed Test Procedures

local t-Test, 3-Slope Test

$H_0$ : slope equal to zero

Change Point Estimation

$H_0$ :  $E(x_i) = \text{const}$

Mann-Kendall

sign-test,  $H_0$ : no trend



## Local t-test

- $(\hat{\alpha}_l, \hat{\beta}_l) = \arg \min_{(\alpha_l, \beta_l)} \sum_{i=1}^n (y_i - \alpha_l - \beta_l x_i)^2 I(x_i \in L_l)$   
where  $L_l = [x_0 + (k-1) \cdot \Delta, x_0 + (k-1) \cdot \Delta + l], l = 1 \dots n/l$   
and  $\Delta$  indicates the step size
- no (local) linear trend,  $H_0: \beta_l = 0$
- t-test statistics  $t_l = \frac{\hat{\beta}_l}{SE(\hat{\beta}_l)}$



## Local t-Test

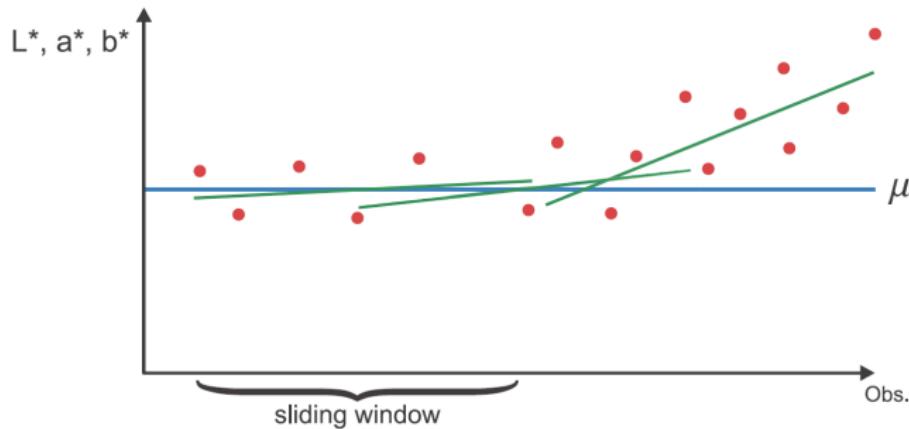


Figure 4: Schematic chart of local t-test



## Change Point Estimation

$$H_0 : \mu_1 = \mu_2 \quad H_1 : \mu_1 \neq \mu_2$$

The different  $\mu$ -values are computed as  $\widehat{\mu}_j = n^{-1} \sum_{i \in I_j} y_i$ , the resulting test statistics is asymptotically  $\chi^2$ -distributed with one degree of freedom.

$$\frac{n_1 \cdot n_2}{n_1 + n_2} \frac{(\widehat{\mu}_1 - \widehat{\mu}_2)^2}{\tilde{\sigma}^2} \xrightarrow{\mathcal{L}} \chi^2(1)$$

$\tilde{\sigma}^2$  calculated by non-parametric smoothing.



## Non-Parametric Smoothing

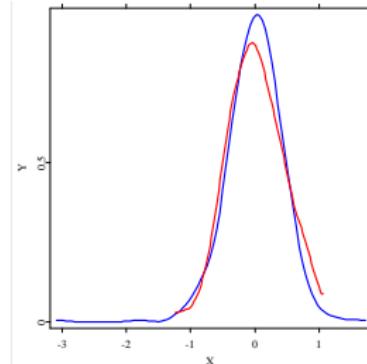
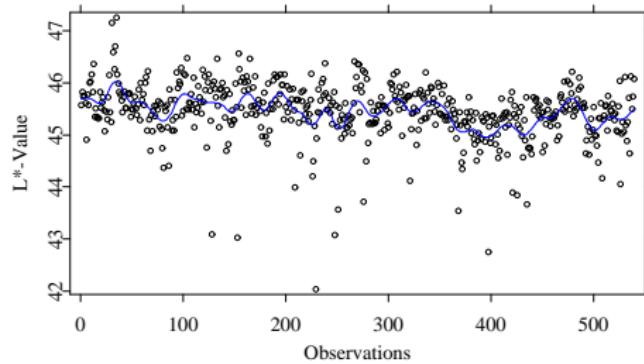


Figure 5: data with smoothed curve and distributions of residuals (blue) and fitted normal distribution pdf (red)



## Mann-Kendall Test

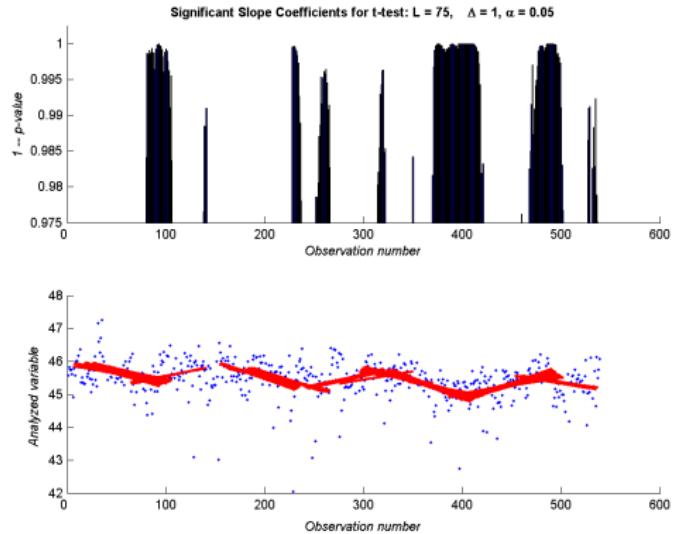
$$S = \sum_{i=2}^n \sum_{j=1}^{i-1} \text{sign}(X_i - X_j)$$

Under the null hypothesis of no trend the test statistics follows approximatively a standard normal distribution.

$$\frac{S}{\sqrt{\frac{n(n-1)(2n+5)}{18}}} \xrightarrow{\mathcal{L}} N(0, 1)$$

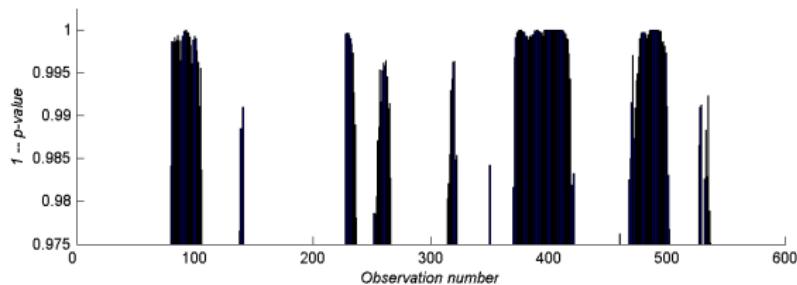


## Example



## Integration Measure

Idea: Look at clusters of bars

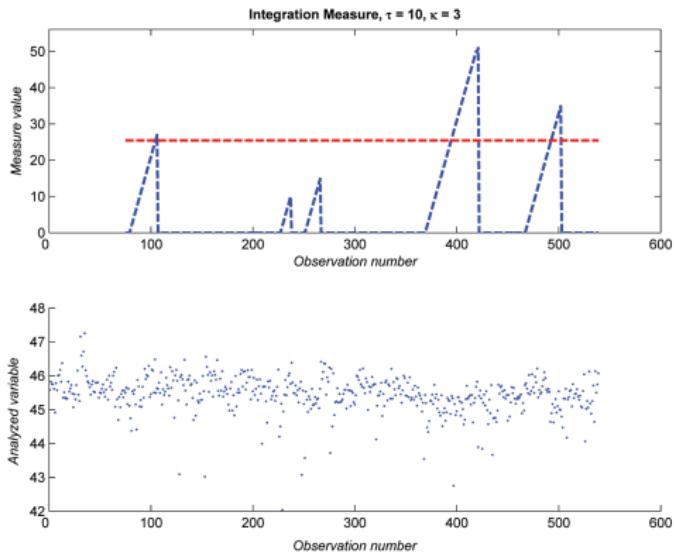


Control false alarms and gaps

- $\tau$  number of subsequent bars necessary to start one integral
- $\kappa$  number of allowed gaps between two groups of bars



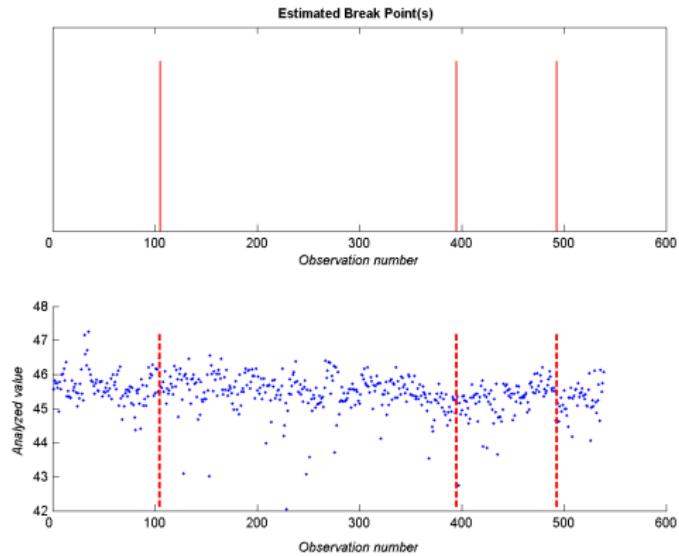
## Example



Trend Detection



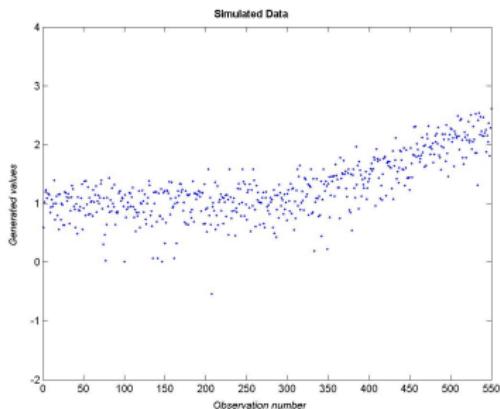
## Example



Trend Detection



## Selection of $\beta$ , $\tau$ and $\kappa$



- regression line with trend at 300 observations
- quasi-residuals calculated by non-parametric regression
- 50 simulation runs for each parameter setting



# Results

$\tau$	$\kappa$	Mean delay (5 tests)						
		$I$	25		50		75	
		3	5	3	5	3	5	
3	3	211	211	213	213	213	213	
5	5	131	131	132	132	132	132	
10	10	71	71	72	72	72	72	

Table 1: Mean delays of all tests for  $\gamma = 2.86^\circ$  and  $\Delta = 5$



## Conclusion

- choice of  $l$ ,  $\kappa$  and  $\tau$
- comparison of tests:
  - local t-test outperforms Mann-Kendall
  - change point estimation performs well
- threshold value to be adjusted by further analysis

Thank You for listening!

