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The control of the mass of an aggregates deliveries and its impact on the effectiveness of earthworks execution

Key words: aggregate supplies control, earthworks effectiveness, statistical weight control, determination of the sample size, determination the mean on the basis of a sample, the effectiveness of aggregate supplies

Introduction

The construction project is considered as successful as it finishes within planned schedule and budget (Skorupka, Kuchta & Górski, 2012). Particularly in case of linear structures and water and sewage networks, the bulk materials are among the most important positions in the project budget, which can reach even 20% of total costs (Ośrodek Wdrożeń Ekonomiczno-Organizacyjnych Budownictwa “Promocja”, 2015). Controlling the deliveries of aggregates under construction conditions is problematic due to the difficulty of assessing the quantity of products that are not packaged in countable units. Human resources problems of these days, as well as shortening construction works time, often makes precise control

of the weight of all transports impossible. Due to that facts, inspection of transports is usually limited only to visual assessment and a control of documents. The lack of reliable control of deliveries can effect in a lower transport efficiency and large discrepancies between ordered and delivered amount of aggregates. Lower transport efficiency contributes to its cost overestimation (Sobotka, Radziszewska-Zielina, Plebankiewicz & Kowalik, 2014) and environmental pollution (Pawłowska, 2018).

The profit is necessary for contractors business. It allows to survive during economic downturn, as well as develop while economics is in excellent condition (Anysz, 2017). Taking into account the average gross profit of construction companies of 6–7% (Deloitte, 2017), such discrepancies can influence not only project success but also contractor financial condition.

The weight control of all incoming (and next, outgoing) dump trucks is affordable only for a limited number of

companies. The easiest method of controlling random trucks based on determining the mean of transported mass can be not enough effective, as it does not answer the question what is the optimal number of samples to be inspected. It does not answer what is the estimation error in that method as well, what can be the reason for questioning its result by a supplier.

The research aims to propose a tool, which allows pointing the optimal number of trucks to be inspected. The method based on statistics allows to determine the average weight of the transport with specified probability and estimation error, so the total weight of supplied goods can be estimated too.

Research method

Usually, the amount of loose material on a truck is estimated on basis of a loading machine parameter (digger, loader, belt conveyor) e.g. a bucket volume. It happens very rarely to control the mass of deliveries on construction site, especially if linear construction works are considered. The visual method of evaluating the loose material volume is prone to large discrepancies as:

- natural materials are considered, which are not homogenous;
- a differential in fractions of the material at respective depths of the spoil may occur;
- variable weather conditions cause the differences in loose materials density.

The population is defined as a number of homogenous dump trucks entering a building site. The weight of

aggregates delivered in a single turn can be considered as normally distributed on following conditions:

- all means of transport have the same capacity;
- all deliveries are made by single company;
- the loading method is the same and deliveries are made from the same mine;
- all machine operators involved in the delivery process (loading, transporting, acceptance) have clear intentions.

All conditions specified above have to be met, otherwise, deliveries have to be divided into separate, smaller homogenous populations, which requires individual calculations for them. These conditions cause, that even on large construction sites the homogenous populations might not be significant.

Depending on quantity of aggregates to be delivered, construction site surrounding, distance to loading place and logistics on construction site, the load capacity of typically used means of transport used for aggregates vary from 8 up to 26 t. Deliveries of underloaded trucks result in a lower total weight of material delivered, lower project profitability and lower transport efficiency. Deliveries of overloaded trucks are desirable for a contractor but may result in penalties for exceeding capacity and decreases supplier efficiency.

Vehicles with capacity of 18 t have been used for calculation. To check versatility of the tool, 3 types of construction site have been analyzed (Table 1).

Authors expect the large standard deviation under construction conditions, even if considering of carelessness and

TABLE 1. Aggregates deliveries, analyzed in the research

Length of roadworks [m]	Width of roadworks [m]	Thickness of aggregates layer [m]	Bulk density of aggregates [kg·m ⁻³]	Compactness	Estimated quantity of 18-tons transport means
1 800	20	0.15	1.65	0.98	500
5 000					1 400
10 000					2 800

unclear intentions of the supplier are omitted. Data rounded to 10 kg for each construction sites have been drawn using 3 methods:

- supplier S1 – random data from population with standard distribution ($\sigma = 240$ kg, $\mu = 17,950$ kg);
- supplier S2 – random data from population with standard distribution ($\sigma = 480$ kg, $\mu = 17,750$ kg);
- supplier S3 – random data, drawn from intervals (0.93·18,000; 18,000) and (1,800; 1.03·18,000) with 37 and 63% probability respectively.

The process of determining the required deliveries to be inspected have to be divided into 3 steps. In described case standard deviation of population (σ) and mean of population (μ) is unknown. Thus in the first step sample standard deviation (S) and sample mean (\bar{X}) have to be determined after initial testing of at least 30 samples. On the basis of S and \bar{X} obtained from the test, in the second step the rest of required trucks to be inspected have to be determined under assumed confidence level and estimation error conditions. The last step includes controlling the appointed dump trucks. On the basis of this result, the mean and estimated sum of deliveries for the whole population have to be calculated.

Due to high restrictions on the homogeneity of the population authors decid-

ed to include finite-population correction factor in confidence level formula.

The finite-population correction factor (fpc) is expressed as follows (Aczel, 1993):

$$fpc = \sqrt{\frac{N-n}{N-1}} \tag{1}$$

where:

N – population size;

n – sample size.

The confidence interval for a mean with unknown standard deviation (Aczel, 1993) with finite-population correction factor (1) is expressed as follows:

$$P\left\{\bar{X} - t_{\frac{\alpha}{2}} \frac{S}{\sqrt{n}} \sqrt{\frac{N-n}{N-1}}; \bar{X} + t_{\frac{\alpha}{2}} \frac{S}{\sqrt{n}} \sqrt{\frac{N-n}{N-1}}\right\} = 1 - \alpha \tag{2}$$

where:

\bar{X} – mean resulted from initial testing;

S – standard deviation of initially tested samples;

$t_{\frac{\alpha}{2}}$ – critical value of t -distribution for assumed confidence level $1 - \alpha$ (the recommended value – 95%), for $n - 1$ degrees of freedom).

To limit possible error to a specific level, authors decided to introduce maximum estimation error, which allows lim-

iting the inaccuracy of the mean to value defined in kilograms.

Due to (2) the estimation error (τ) satisfies the following inequality:

$$\frac{t_\alpha}{2} \frac{S}{\sqrt{n}} \sqrt{\frac{N-n}{N-1}} \leq \tau \quad (3)$$

From (3) the minimum number of required samples to be tested:

$$n \geq \frac{N \cdot t_\alpha^2 \cdot S^2}{\tau^2 \cdot (N-1) + t_\alpha^2 \cdot S^2} \quad (4)$$

The estimated mean of the population (with assumed estimation error and confidence level):

$$\bar{X} = \frac{\sum_{i=1}^n m_i}{n} \pm \tau \quad (5)$$

where:

m_i – mass of aggregates delivered by single dump truck i ($i = 1, 2, 3, \dots, n$) [kg];

τ – estimation error of the mean [kg].

Example calculation based on Excel tool. Authors have developed a tool, which uses the equations mentioned above (Fig. 1).

The first sheet “Data” is developed for entering data obtained from initial 30 sample test. Second sheet “Number of transports” is used for determining the required number of dump trucks to be inspected under assumed confidence level and estimation error conditions.

The lower estimation error is assumed, the more dump trucks should be weighted. It is possible to find the optimum estimation error (Fig. 2), minimizing the total cost (c) given by the formula:

$$c(\tau) = \tau \cdot N \cdot p_a + n \cdot p_w \quad (6)$$

where:

p_a – price of aggregates [PLN·kg⁻¹];

n – number of required samples to be tested, determined from (3), depended on τ ;

p_w – price of single truck weighting [PLN].

On the basis of 9 variants described in Table 2 and current prices of loose materials, the optimal estimation error is usually between 30 and 200 kg.

Total amount of deliveries	2800
Confidence level	0,95
Initial sample	30
Mean of initial sample	17798,00
Standard deviation of initial sample	575,78
Estimation error [kg]	100
Number of trucks to control	131

Appoint the number of trucks which should be inspected

FIGURE 1. Screenshot of “Number of transports” sheet used for the required sample calculation

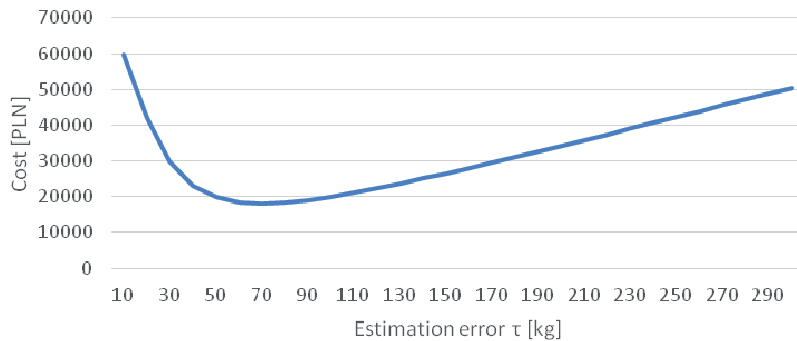


FIGURE 2. The optimum error appointment (supplier S2, $N = 2,800$, $p_a = 60 \text{ PLN} \cdot \tau^{-1}$, $p_w = 25 \text{ PLN}$)

TABLE 2. Analyzed populations (unknown in real case)

Supplier	Population			
	quantity [pcs]	μ [kg]	σ [kg]	sum [t]
S1	500	17 939	236.92	8 970
	1 400	17 946	237.16	25 124
	2 800	17 944	237.42	50 243
S2	500	17 765	506.52	8 883
	1 400	17 749	475.81	24 848
	2 800	17 758	477.46	49 722
S3	500	17 675	528.40	8 838
	1 400	17 687	533.76	24 761
	2 800	17 696	535.75	49 550

The last sheet “The result” is to enter the missing data (the rest of transports to control) and estimate the total sum of deliveries. During the research, 9 variants presented in Table 2 have been analyzed. The results of tests are shown in Table 3.

Results

The initial control of transport mass, based on 30-element sample, allows evaluating the quality of supplier. Depending on the standard deviation and

the mean it is possible to conclude if the supplier can be considered as reliable. Then earthworks quantity survey done on the basis of deliveries total volume should confirm the result. The high value of standard deviation can also lead to search for reasons of heterogeneity of populations. It can be a milestone, after which decision on further cooperation or keeping contract terms unchanged can be made. Nevertheless, the supplier assessment based on initial test results distribution can be confusing. Despite the fact, that deliveries from supplier 3

TABLE 3. The results of the test

Supplier	Initial test		Appointed test					
	\bar{X} [kg]	S [kg]	trucks [pcs]	τ [kg]	\bar{X} [kg]	S [kg]	estimated sum [t]	τ related to population [t]
S1	17 897	259.68	131	40	17 939	244.82	8 970	20
			157		17 945	240.76	25 123	56
			166		17 945	236.52	50 246	112
S2	17 798	575.78	109	100	17 764	573.80	8 882	50
			127		17 738	569.58	24 833	140
			133		17 750	571.38	49 700	280
S3	17 570	502.85	88	100	17 736	511.54	8 868	50
			99		17 708	506.56	24 791	140
			102		17 707	499.89	49 580	280

are not normally distributed, the visual evaluation based on 30 tests can lead to that conclusion.

As the Kolmogorov–Smirnov test cannot be applied according to unknown distribution parameters of the population (Kot, Jakubowski & Sokołowski, 2011), Shapiro–Wilk test (Rabiej, 2012) was applied. It is widely used test (Słowik & Rogalska, 2013; Kępczak & Woyciechowski, 2015), as it is recognized as a powerful one (Kot et al., 2011). For all suppliers, the test shows $p > 0.05$. Together with visual evaluation of histograms (Statsoft, 2005), samples taken from each supplier can be treated as taken from the population of the normal distribution. The histogram of the sample taken from supplier 2 seems the weakest matching the curve but on the still moderate level of matching the normal distribution (Fig. 3).

Results of test mentioned above are summarized in Table 4. The same evaluation was done for the samples of the size calculated for the population 500, 1,400

and 2,800 deliveries, for each supplier. It is to emphasize that, Shapiro–Wilk test and visual evaluation excluded supplier 3 from normal distributions (for the sample size bigger than 30 elements).

At the same time fitting level (visual and p value) for suppliers 1 and 2 has been improved much (Fig. 4.)

Hypothesis testing about the population means

When the standard deviation of the population is unknown, but it is known for the sample, the t-Student test can be applied. The other assumption, claimed by many statisticians (Aczel, 1993), is normality of the population. The hypothesis to check is: if the mean in the population (μ) is higher than mean from the sample increased by assumed error (so-called one side test). For instance, for the supplier 1 and population 500 deliveries, we can write:

$$H_0: \mu \geq 17,979 \tag{7}$$

$$H_1: \mu < 17,979 \tag{8}$$

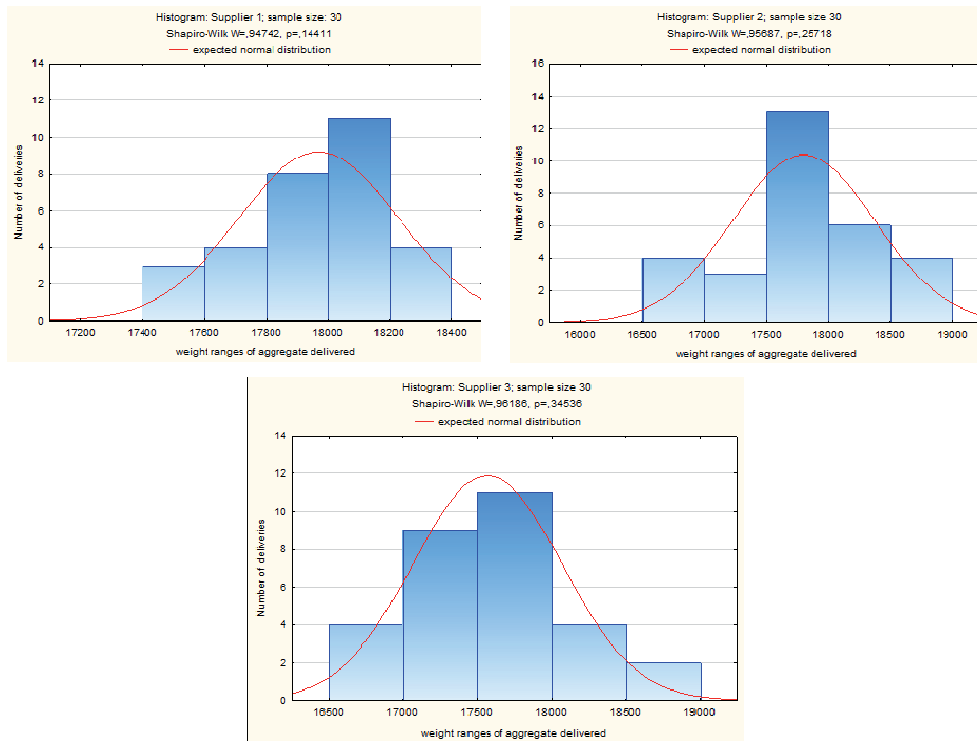


FIGURE 3. Histograms and fitting the normal distributions of 30-element sample for each supplier

TABLE 4. The evaluation of initial tests distribution

Supplier	sample size	Shapiro–Wilk tests		Visual evaluation of conformity with the normal distribution	Evaluation of conformity with the normal distribution
		W	p		
1	30	0.947	0.144 > 0.05	moderate	yes
2	30	0.957	0.257 > 0.05	moderate	yes
3	30	0.962	0.345 > 0.05	moderate	yes

The alternative hypothesis is that the average weight of a single delivery in the whole population of them is lower than 17,979 kg. The t value can be calculated using the formula:

$$t = \frac{\bar{X} - \mu}{S/\sqrt{n}} \quad (9)$$

As the t value calculated for all of the suppliers is lower than t read from tables, alternative hypothesis H_1 should be accepted. It means that the mean weight in the all 9 cases, with 95% confidence is lower than the sample mean plus assumed error. It is to remember, that calculating the size of sample, the levels of

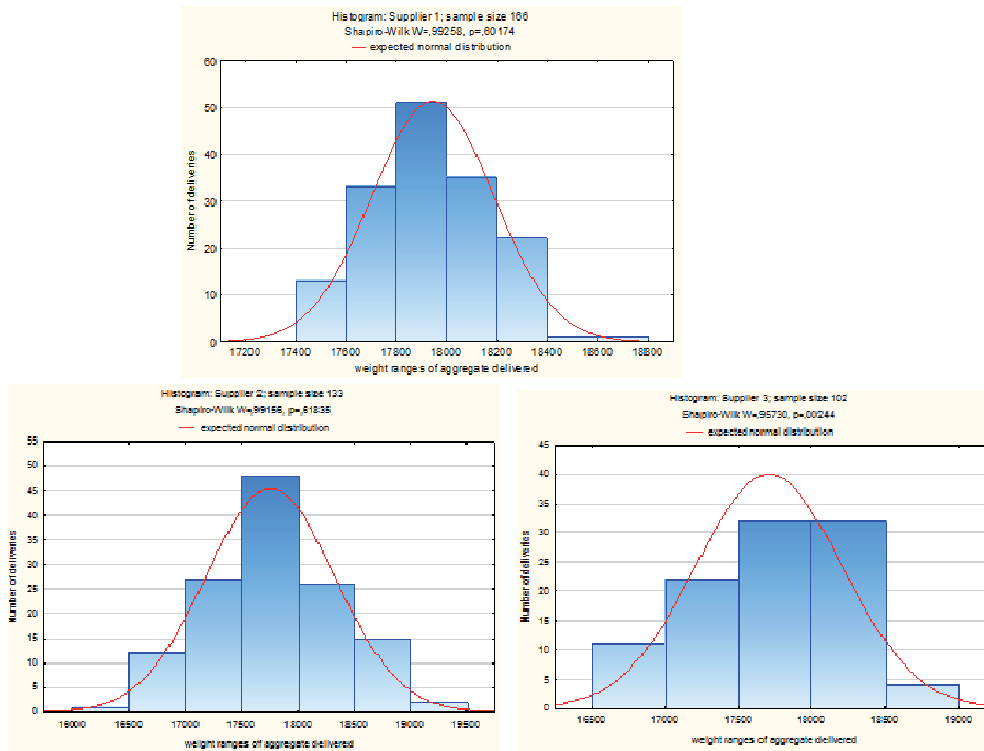


FIGURE 4. Histograms and fitting the normal distributions of samples relevant for 2,800 population for each supplier

TABLE 5. Results of t-test (tables of $t_{0,05}$ values for one side test taken from (Lissowski, Haman & Jasiński, 2011)

Item	Supplier 1			Supplier 2			Supplier 3		
	500	1 400	2 800	500	1 400	2 800	500	1 400	2 800
Population	500	1 400	2 800	500	1 400	2 800	500	1 400	2 800
Sample size	131	157	166	109	127	133	88	99	102
<i>df</i>	130	156	165	108	126	132	87	98	101
<i>t</i> read from tables	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66
<i>t</i> calculated	-1.870	-2.082	-2.179	-1.820	-1.979	-2.018	-1.834	-1.964	-2.020

error were assumed (40 kg for supplier 1 and 100 kg for suppliers 2 and 3). As it is shown in Table 6 that there is no base to reject another hypothesis H_0 that mean of the population is equal or higher than the sample mean minus assumed error for all suppliers.

It was confirmed by hypothesis testing that using the created excel tool for estimating the population mean on the assumed confidence level 95% were calculated correctly.

TABLE 6. Results of t-test for $H_0: \mu \geq$ the sample mean minus assumed error, calculated for all the suppliers

Item	Supplier 1			Supplier 2			Supplier 3		
	500	1 400	2 800	500	1 400	2 800	500	1 400	2 800
Population	500	1 400	2 800	500	1 400	2 800	500	1 400	2 800
Sample size	131	157	166	109	127	133	88	99	102
<i>df</i>	130	156	165	108	126	132	87	98	101
<i>t</i> read from tables	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66	-1.66
<i>t</i> calculated	1.870	2.082	2.179	1.820	1.979	2.018	1.834	1.964	2.020

Summary and discussion

The results of calculations show the correctness of assumed method for calculation (Table 3). The mean weight of transport for suppliers 1 and 2 is included within assumed accuracy.

Although supplies S3 cannot be considered as normally distributed, the mean weight of the transport is determined correctly.

The distribution of final testing results can answer much more questions. If

distribution cannot be evaluated as normal, either the population was incorrectly assumed as homogenous, or the mass of transports have been manipulated on purpose. To evaluate the finite-population factor influence, 10,000 simulation of deliveries for each nine variants have been made. The result of the simulation is presented in Table 7.

The last column of Table 7 shows the percent of deliveries, where the estimated mean on the basis of the appointed

TABLE 7. Simulation of 10,000 deliveries per 9 variants

Supplier	Population		Appointed test		
	quantity [pcs]	μ [kg]	trucks averaged [pcs]	\bar{X} [kg]	deliveries with estimation error exceeded [%]
S1	500	17 950.0	114	17 950.0	5.03
	1 400	17 950.2	134	17 950.0	5.00
	2 800	17 949.4	141	17 950.0	5.05
	infinity*	17 950.3	150	17 950.1	2.35
S2	500	17 749.8	79	17 749.7	4.99
	1 400	17 749.5	89	17 750.0	4.93
	2 800	17 749.8	92	17 750.3	4.97
	infinity*	17 749.8	95	17 749.9	3.57
S3	500	17 704.2	95	17 702.9	4.52
	1 400	17 704.3	108	17 702.9	5.07
	2 800	17 704.1	112	17 703.0	4.80
	infinity*	17 704.4	117	17 703.0	2.67

*Calculation without finite-population factor.

number of weight controls have exceeded the estimation error.

In cases where finite-population factor have been used, this value is very close to assumed significance level $\alpha = 5\%$. This proves that excluding the finite-population factor in calculation results in the non-optimal appointed number of controls, as deliveries which exceed the estimation error not correspond with assumed significance level. What is surprising, although the distribution of supplies S3 is not normal, the mean weight calculated on the basis of appointed trucks number and significance level turned out to be correct.

Conclusions

The insufficient quality control is an important and up-to-date problem of construction projects (Deszcz, 2017). The statistics can be used for determining the optimal number of trucks, which mass should be verified to estimate the total mass of supplied aggregates. What is more, it allows optimizing the controlling costs of deliveries. The mean appointed on the basis of statistical test can be used to evaluate the total sum of deliveries. The proposed method can contribute to the project success as it reduces the risk of payment for unrealized deliveries.

Nevertheless before initial testing a research should be done, in order to find the homogenous population of transports. The calculation should be supplemented by visual assessment of distribution of controlled transports.

The contract terms can affect much the project success (Czaczkowski, 2013).

The proposed tool and its assumptions can be the base of a settlement between contractor and aggregates supplier described in contract terms. The method can be used as well by a supplier to optimize the usage of transport means. Many surveyors proved, that proper choice of contractor or subcontractor affects the effectiveness of the project (Leśniak, Plebankiewicz & Zima, 2012; Ibadov, 2015; Biruk, Jaśkowski & Czarnigowska, 2017). The described case proves, that also the proper choice of materials supplier can influence the cost of the project, so affects its effectiveness as well.

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Summary

The control of the mass of an aggregates deliveries and its impact on the effectiveness of earthworks execution. The paper presents the method and tool which can be used to control the mass of aggregates deliveries under construction site conditions. The method based on statistics allows determining the optimal quantity of transports to be inspected, required to estimate the total sum of loose materials deliveries assuming estimation error and confidence level conditions. Inspection based on described method allows to improve the effectiveness of earthworks execution as well as gives the possibility to evaluate the quality of the supplier.

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