Optimization of the compressive strength of Palm Kernel Shell lightweight Aggregate concrete Using Scheffe`s Modeling Theory

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ABSTRACT: The environmental consequence of natural aggregate exploitation and Agro-base waste management especially open-air incineration is immense. In the construction industry, the role played by structural light weight concrete cannot be quantified enough especially in the case of high-rise buildings. The light weight concrete can be produced only by using light weight aggregates such as palm kernel shells in concrete mixture, and the introduction of this light weight aggregate in concrete is cost saving in construction. In this research, Henry Scheffe's regression theory was used to develop a mathematical model to predict and optimize the compressive strength of palm kernel shells aggregate concrete. A total of seventy (70) cubes were produced for the experimental process for the thirty-five (35) points used to determine the coefficients of the model, while fifteen mix ratios consisting of a total of thirty (30) cubes were used for validation of the model. The mathematical model results obtained conform favourably with the experimental results. Validation and test of adequacy of the model was based on statistical analysis for the control points using ANOVA where the adequacy was tested using student t-test and fisher f-test at 95% confidence level and found to be adequate.A compressive strength of 5.45N/mm² corresponding to mix ratio of 0.5875: 1.0: 1.73755: 2.4625 for water, cement, fine aggregates and PKS respectively, was predicted by the model and laboratory strength of 3.80Mpa all of which are less than the minimum of 17.2Mpa specified for structural lightweight concrete. A wolframl computer program was coded to obtain the optimized compressive strength of the palm kernel shells concrete.

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I. INTRODUCTION

Concrete has wide area of application in Civil Engineering and building works and due to urbanization especially in developing countries, the use of concrete product has continued to be on its increase; leading to increased depletion of natural resources and thereby distorting the ecological balance. [1]. Concrete is a very variable material having a wide range of strength and concrete generally increases its strength with age. Concrete is the most widely used construction material worldwide, which is due to its versatility, strength, durability and ease to place into forms and shapes [2,3,4]. Concrete is composed principally of aggregate, a Portland or blended cement, and water, and may contain other cementitious materials and/or chemical admixtures. Chemical admixtures used to accelerate, retard, improve workability, reduce mixing water requirements, increase strength, or alter other properties of the concrete. [5,6]. Concrete as a major construction material has a high demand leading to decrease in granite and gravel deposit hence their scarcity and expensive in cost where available[7]. To mitigate against the continuously increasing demand for low cost and environmentally friendly construction materials, while strengthening economic growth and competitiveness, agricultural waste can be used as replacement material in construction industry especially where this waste is in abundant. [8].

Professionals in the built environment has for long period made great efforts towards reducing the environmental impact of the construction process through the use of alternative construction materials such as Agro-based waste, industrial based waste, etc, in recycled formed of which Palm Kernel Shell is prominent.

The management and environmental menace associated with agricultural and industrial waste has been of challenges to engineers and experts in different engineering related fields especially the environmental nuisance caused by the open field deposit incineration of such waste as palm kernel shells among others [9].

Concrete mix design is the technique of aptlychoosing the proportions of constituent materials such as cement, water, fine aggregates, and coarse aggregate and admixtures where conceivable so as to produce concrete satisfying all the required properties for minimum cost. Essentially, two treasured conditions to attain

economy in mix design process are the use of locally available materials and adoption of less restraining specification requirements. [10, 11].

With the application of concrete in the construction industries, a workable design mix through optimization yielding conceivable mixture combination of components for the required maximum strength has become imperative. Optimization of the mix proportion in concrete production could beneficially impact on the construction project cost than when trial mix is unceasingly employed, which reduces the waste of individual component materials of concrete compared with experiential methods of trail mix.[12]. The task of concrete mix optimization is to estimate different concrete composition with different composition of aggregate, to choose the best alternatives of mix by comparing their economical and mechanical properties, includingmaterial durability [13].

Simplex is the structural depiction (shape) of lines or planes joining presumed positions of the constituent materials (atoms) of the mixture and are equidistant from each other. The atoms are the constituent components of the mixture and according to Henry Scheffe, the property studied in the mixture depends on the component proportions and not their quantities[14]. The choice of the suitable mixture design entailstaking account of some points; such as the number of factors and interactions to be studied, the complexity of each design, the statistical validity and effectiveness of each design, and the ease of execution and cost, time constraints associated with each design. The most recurrently used mixture design types are the simplex lattice design and simplex- centroid design [15].

Scheffe's optimization theory is used to optimize compressive strength of four component four-degree (4, 4) polynomial model at 28 days curing when granite is 100% replaced by PKS as coarse aggregate.

The compressive strength is the most common measure for judging the quality of concrete and the characteristics of concrete based on the 28day cube strength. It is what knowing that the concrete strength is normally specified in terms of characteristic strength at a particular given age of the concrete. (this is crushing strength of standard 150mm cubes at an age of 28 days after mixing) (Kong & Evans, 1986)[16]. Compressive and tensile strengths of concrete are important parameters utilized in the analysis and design of concrete members. (Mutiu et`al, 2017). [17]

The four-dimensional factor spaces for four componentsfour-degree polynomial regressions for the simplex design are presented in figure below.

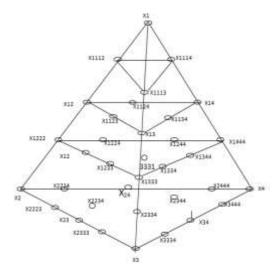


Fig.1: A four-dimensional factor space of four components

II. METHODOLOGY

MATHEMATICAL MODELING AND FORMATION

In concrete mix design, mathematical modelling has various applications with predictive models such as Henry Scheffe's mixture design model to predict concrete properties like compressive, tensile and flexural strengths being the commonest in application. [18, 19]. Scheffe's simplex model has been applied passably to develop mathematical models for concrete mix to predict and optimize properties of concrete such as the modulus of rupture and flexural strength etc, [20, 21, 22]. The recordfrequently used material in the construction industry is concrete and it is required that the strength at 28 days curing be tested for fulfilment of at least a minimum strength of 75% before being put to use. Optimization of the concrete mixture design is a course of

searching for a mixture for which the sum of the costs of the ingredients is lowest, yet satisfying the obligatory performance of concrete, such as workability, strength and durability [23].

As the worth of palm kernel shells in concrete increases, the specific area increases requiring more cement paste for proper bonding because strength requires virtuous bonding of the aggregate and cement. Therefore, as bonding reduces with increase in replacement of palm kernel shells, the compressive strength reduces [7].

INTRODUCTION TO FACTOR SPACE IN SIMPLEX DESIGN

The Scheffe's simplex lattice method is a single step multiple comparison techniqueemploying the use of a single regression polynomial to compare all the constituents in a single step engendering the value of the objective function [24]. Scheffe's method stretchespure understanding of how proportioning the constituents of the concrete affect the engineering behaviours. In simplex lattice method to designing experiment to bout mixture problems regarding component property diagrams, the property studied is assumed a continuous function of certain opinions and with a sufficient accuracy it can be approximated with a polynomial. For multicomponents systems the use of experimental design methodologies substantially reduces the volume of an experimental effort [24][25]. Henry Scheffe developed a theory for experiments with mixture of which the property studied depends on the proportions of the components present and not on the quantity of the mixture. Scheffe showed that if q represents the number of constituent components of the mixture, the space of the variables known also as the factor space is a (q - 1) dimensional simplex lattice. The composition may be expressed as molar, weight, or volume fraction or percentage [26]. Simplex is the structural depiction (shape) of lines or planes joining assumed positions of the constituent materials (atoms) of the mixture and are equidistant from each other where the atoms are the constituent components of the mixture. [27]. The study of the relationship between the compressive strength of concrete and the proportion of water, cement, fine and a coarse aggregate for normal concrete mixture is a good example. Okaforet'al2009[28], has defined simplex as a convex polyhedron with (K+1) vertices produced by K intersecting hyper planes in k- dimensional space. The hyper planes refer to any co-ordinate system above 3- dimensions. This therefore gives a simplex of a mixture of four components and the simplex lattice of this four-component mixture is a 3- dimensional solid equilateral tetrahedron while 2- dimensional regular simplex is referred to as equilateral triangle. The factor space is a regular (q-1) dimensional simplex, and for the whole factor space of a mix design with evenly spaced distribution of points over the factor space is {q, m}[25]. Kenneth2019[29], Scheffe's lattice design provides a uniform scatter of points over a (q-1) simplex where the points form a (q-1) lattice on the simplex and q is the number of mixture components while `n` is the degree of the polynomial. Therefore, for binary system (q = 2), the required simplex is a straight line, for (q = 3), the required simplex is an equilateral triangle while for (q = 4), it is a regular tetrahedron.

SCHEFFE'S FACTOR SPACE

The strength of concrete depends on the adequate proportioning of its ingredients (components), and Scheffe developed an optimization theory that was used to optimize the strength of concrete. The Scheffe's optimization theory can be used to analyse and predict possible mix proportions of concrete ingredient that can estimate/predict a desired concrete strength. The property studied in the mixture depends on the component proportions and not their quantities[**30**]. H. Scheffe 1958[**31**], stated that the property (response) of the mixture is assumed to be a real-value function on a simplex and introduced an appropriate form of polynomial regression model. The polynomial function of degree, n, in the q variables $x_1, x_2, ..., x_q$ must subject to the constraint that;

$$\sum_{i=1}^{q} x_{i=1}$$

(3)

(2)

(1)

Let
$$n = 1$$
: $f(x) \sum_{i=1}^{q} \beta_i x_i$
Let $n = 2$: $f(x) \sum_{i=1}^{q} \beta_i x_i + \sum_{i=1}^{q} \beta_{ij} x_i x_j$

Let
$$n = 3$$
: $f(x) \sum_{i=1}^{q} \beta_i x_i + \sum_{1 \le i \le q}^{1 \le i \le q} \beta_{ij} x_i x_j + \sum_{1 \le i \le j \le k \le q} (\beta_{iij} x_i^2 + \beta_{iij} x_i x_i x_k)$ (4)

NUMBER OF COEFFICIENT (4, 4)

q = 4, m = 4

$$N = \frac{(q+m-1)!}{m! (q+m-1-m)!}, \quad N = \frac{(4+4-1)!}{4! (4-1)!}, \quad N = \frac{7!}{4! \, 3!} = 35$$
(5)

FOUR COMPONENT FACTOR SPACE

Infour-component factor space, there are thirty-five (35) design points with the first four pseudo components located at the vertices of the quartic simplex; while the remaining mix ratios are located at the mid-points of the lines joining the vertices of the simplex as presented below.

A1[1,0,0,0]	A2[0,1,0,0]	A3[0,0,1,0]	A4[0,0,0,1]	A12[0.5,0.5,0,0]
A13[0.5,0,0.5,0]	A14[0.5,0,0,0.5]	A23[0,0.5,0.5,0]	A24[0,0.5, 0, 0.5,]	A34[0,0.0.5,0.5]
A1112[0.75, 0.25, 0, 0]	A1113 [0.75, 0, 0.25,	A1114 [0.75, 0, 0,	A2223 [0, 0.75, 0.25,	A3334 [0, 0, 0.75, 0.25]
66 80	0]	0.25]	0]	2011 20
A1222 [0.25, 0.75,0,0]	A3331 [0.25, 0, 0.75,	A1444 [0.25, 0, 0,	A1233[0.25,0.25,0.5,0]	A1344[0.25,0,0.25,0.5]
ST 8 3454	0]	0.75]	제 - 2019년 1월 1941	
A1123[0.5,0.25,0.25,0]	A1134[0.5,0,0.25,0.25]	A2344[0,0.25,0.25,0.5]	A2234[0,0.5,0.25,0.25]	A1124[0.5,0.25,0,0.25]
A1244[0.25,0.25,0,0.5]	A2224[0,0.75,0,0.25]	A2333[0,0.25,0.75,0]	A2444[0,0.25,0,0.75]	A3444[0,0,0.25,0.75]
A2334[0,0.25,0.5,0.25]	A1223[0.25,0.5,0.25,0]	A1334[0.25,0,0.5,0.25]	A1224[0.25,0.5,0,0.25]	A1234[0.25,0.25,0.25,0.25]

RESPONSES

The constituent elements for a normal concrete mixture are; water, cement, fine and coarse aggregates which thus give a simplex of a mixture of four components. The simplex lattice of this four-component mixture is a three- dimensional solid equilateral tetrahedron. The mixture components here are subjected to Scheffe's constraint that the sum of all the components must be equal to unity (one). If q is the total components and xi is the proportion of the components of the ith component in the mixture, is such that;

$$X_i \ge 0 \ (i = 1, 2 - - q) \text{ and } 0 \le X_i \le 1$$
 (6)

Thus, the sum of the component constituent proportion is whole unity (one), and

$$X_1 + X_2 + X_3 + X_4 = 1 \text{ or, that } \sum X_i - 1 = 0$$
(7)

$$Y = b_0 + \sum b_i x_i + \sum b_{ij} x_i x_j + \sum b_{ijk} x_i x_j x_k + \dots + \sum b_{i1}, i2...i_n x_{i1} x_{i2} x_{in}$$
(8)

with, $1 \le i \le q$, $1 \le i \le j \le q$, $1 \le i \le j \le k \le q$, $1 \le i 1 \le i 1 - -- \le in \le q$ and bo is the constant coefficient. In a {4, 4} lattice which is a four components and four-degree polynomial lattice there exists five levels desired as follows: $X_4 = 0, \frac{1}{4}, \frac{2}{4}, \frac{3}{4}, \frac{4}{4}$, equally expressed as 0, 0.25, 0.5, 0.75, 1.0

The general form of the Polynomial is given as,

$$\hat{\mathbf{y}} = \mathbf{b}_0 + \mathbf{b}_1 \mathbf{x}_1 + \mathbf{b}_2 \mathbf{x}_2 + \mathbf{b}_3 \mathbf{x}_3 + \mathbf{b}_{12} \mathbf{x}_1 \mathbf{x}_2 + \mathbf{b}_{13} \mathbf{x}_1 \mathbf{x}_3 + \mathbf{b}_{23} \mathbf{x}_2 \mathbf{x}_3 + \mathbf{b}_{11} \mathbf{x}_1^2 + \mathbf{b}_{22} \mathbf{x}_2^2 + \mathbf{b}_{33} \mathbf{x}_3^2$$
(9)

While,

$$X1+X2+X3 = 1$$
 (10)

And,

$$b_0 x_1 + b_0 x_2 + b_0 x_3 = b_0 \tag{11}$$

Multiplying eq. (10) in success by X1+X2+X3, $x_1^2 = x_1 - x_1 x_2 - x_1 x_3$

$$x_2^2 = x_2 - x_1 x_2 - x_2 x_3$$
 (12)

 $x_3^2 = x_3 - x_1 x_3 - x_2 x_3$ Substituting eq. (11) and (12) into eq. (9) and after essential transformation will yield $\dot{\mathbf{y}} = (\mathbf{b}_0 + \mathbf{b}_1 + \mathbf{b}_{11}) \mathbf{x}_1 + (\mathbf{b}_0 + \mathbf{b}_2 + \mathbf{b}_{22}) \mathbf{x}_2$

+
$$(b_0 + b_3 + b_{33}) x_3 + (b_{12} - b_{11} - b_{22}) x_1 x_2$$

+
$$(b_{13} - b_{11} - b_{33}) x_1 x_3 + (b_{23} - b_{22} - b_{33}) x_2 x_3$$
 (13)

Means that; $\alpha_i = b_0 + b_i + b_{ii}$, and $\alpha_{ij} = b_{ij} - b_{ji}$ 14)

Giving the reduced second-degree Polynomial in three variables

 $\dot{y} = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{23} x_3 x_3$ (15)The general form for the reduced Polynomial of (4, 4) model can be represented below [32] $\dot{y} = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 + \alpha_4 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{23} x_2 x_3 + \alpha_{24} x_2 x_4 + \alpha_{34} x_3 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{23} x_2 x_3 + \alpha_{24} x_2 x_4 + \alpha_{34} x_3 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{23} x_2 x_3 + \alpha_{24} x_2 x_4 + \alpha_{34} x_3 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{13} x_1 x_3 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{14} x_1 x_4 + \alpha_{12} x_1 x_2 + \alpha_{14} x_1 x_4 + \alpha_{14} x_1 x_1 + \alpha_{14} x_1 x_1 + \alpha_{14} x_1 + \alpha_{14} x_1 + \alpha_{14} x_1 + \alpha_$ $+ \lambda_{12} x_1 x_2 (x_1 - x_2) + \lambda_{13} x_1 x_3 (x_1 - x_3) + \lambda_{14} x_1 x_4 (x_1 - x_4) + \lambda_{23} x_2 x_3 (x_2 - x_3) + \lambda_{24} x_2 x_4 (x_2 - x_4)$ $+\lambda_{34}x_{3}x_{4}(x_{3}-x_{4}) + \mu_{12}x_{1}x_{2}(x_{1}-x_{2})^{2} + \mu_{13}x_{1}x_{3}(x_{1}-x_{3})^{2} + \mu_{14}x_{1}x_{4}(x_{1}-x_{4})^{2} + \mu_{23}x_{2}x_{3}(x_{2}-x_{3})^{2}$ $+ \mu_{24}x_2x_4 (x_2 - x_4)^2 + \mu_{34}x_3x_4 (x_3 - x_4)^2 + \alpha_{1123}x_1^2x_2x_3 + \alpha_{1124}x_1^2x_2x_4 + \alpha_{1134}x_1^2x_3x_4 + \alpha_{1223}x_1x_2^2x_3 + \alpha_{1124}x_1^2x_3x_4 + \alpha_{1223}x_1x_2^2x_3 + \alpha_{1124}x_1^2x_3x_4 + \alpha_{1223}x_1x_2^2x_3 + \alpha_{1223}x_2^2x_3 + \alpha_{1223}x_1x_2^2x_3 + \alpha_{1223}x_2^2x_3 + \alpha_{1223}x_2^2x_3 + \alpha_{1223}x_2^2x_3 + \alpha_{1223}x_2^2x_3$ $+ \alpha_{1224}x_1x_2^2x_4 + \alpha_{2234}x_2^2x_3x_4 + \alpha_{13}$

 $\alpha_{1244}x_1x_2x_4^2 + \alpha_{1344}x_1x_3x_4^2 + \alpha_{2344}x_2x_3x_4^2 + \alpha_{1234}x_1x_2x_3x_4 \quad (16)$ To obtain the value of the coefficients, we substitute in succession the coordinates of all the thirty-five points of the design matrix in eq. (16)

The general equations for the coefficients are generated follows:

$$\alpha_i = y_i \tag{17}$$

$$\alpha_{ij} = 4y_{ij} - 2y_i - 2y_j \tag{18}$$

$$\lambda_{ij} = (8/3) \left(-y_i + 2y_{iiij} - 2y_{ijj} + y_j \right)$$
(19)

$$\mu_{ij} = (8/3)(-y_i + 4y_{iiij} - 6y_{ij} + 4y_{ijjj} - y_j)$$
(20)

$$\frac{1}{16/3} - \frac{32(3y_{iijk} - y_{ijjk} - y_{ijjk}) + (63)(0y_i - y_j - y_k) - 10(y_{ij} + y_{ik})}{-(16/3)(5y_{iiij} + 5y_{iiik} + 3y_{ijji} - 3y_{ikkk} - y_{ijjk} - y_{jkkk})}$$
(21)

$$\alpha_{ijjk} = 32(3y_{ijjk} - y_{iijk} - y_{ijjkk}) - (8/3)(6y_j - y_i - y_k) - 16(y_{ij} + y_{jk}) - (16/3)(5y_{iii} + 5y_{iiik} - 3y_{iikk} - y_{iiik}) - (16/3)(5y_{iii} + 5y_{iiik} - 3y_{iikk} - y_{iiik}) - (16/3)(5y_{iii} + 5y_{iiik} - 3y_{iikk} - y_{iikk}) - (16/3)(5y_{iii} + 5y_{iik} - 3y_{iikk} - y_{iikk}) - (16/3)(5y_{iii} + 5y_{iik} - 3y_{iikk} - y_{iikk}) - (16/3)(5y_{iii} + 5y_{iik} - 3y_{iikk} - y_{iikk}) - (16/3)(5y_{iik} - y_{iikk} - y_{iikk} - y_{iikk}) - (16/3)(5y_{iik} - y_{iikk} - y_{iikk} - y_{iikk}) - (16/3)(5y_{iik} - y_{iikk} - y_{ii$$

$$\alpha_{ijkk} = 32(3y_{ijkk} - y_{iijk} - y_{iijk}) + (8/3)(6y_k - y_i - y_j) - 16(y_{ik} + y_{jk})$$
(22)

$$- (16/3)(5y_{ikkk} + 5y_{jkkk} - 3y_{iijk} - 3y_{iijj} - y_{iijj})$$
(23)

 $\alpha_{ijkl} = 256y_{ijkl} - 32(y_{iijk} + y_{iikl} + y_{ijjk} + y_{ijjl} + y_{ijkl} + y_{ijkk} + y_{ikkl} + y_{ijll} + y_{jkll} + y_{ijll} + y_{ijkl} + y_{ikll}) \\ + (32/3)(y_{iiij} + y_{iiik} + y_{iiil} + y_{ijjl} + y_{jijk} + y_{jijl} + y_{ijkk} + y_{jkkk} + y_{jkkk} + y_{kkkl} + y_{illl} + y_{jlll} + y_{klll})$ (24)

ACTUAL COMPONENT AND PSEUDO COMPONENT

Let AZ = AX

(25)

Where; Z represent the actual components and X represents the pseudo components and A is a constant for a four-by-four matrix. The value of the matrix A is obtained from the first four mix ratios with the corresponding pseudo components as;

Z1 [0.65:1.0:2.0:2.85]; Z2 [0.60:1.0:1.75:2.5]; Z3 [0.55:1.0:1.55:2.2]; Z4 [0.70:1.0:2.3:3.25]

Corresponding mix ratios;

 $-22(2\pi)$

X1[1:0:0:0]; X2[0:1:0:0]; X3[0:0:1:0]; X4[0:0:0:1];

The actual mixture components can be determined using the corresponding pseudo components when xi and zi are substituted in eq. (17 - 24)

X1 = fraction of water-cement ratio

X2 =fraction of cement

X3 = fraction of fine aggregate

X4 = fraction of palm kernel shell

[Z1]	1	a ₁₁	a ₁₂	a ₁₃	a ₁₄]	[X1]	
Z1 Z2 Z3 Z4	_	a ₂₁	a ₁₂ a ₂₂ a ₃₂ a ₄₂	a ₂₃	a ₂₄	X2	
Z3	=	a ₃₁	a ₃₂	a ₃₃	a ₃₄	X3	
Z4		a ₄₁	a ₄₂	a ₄₃	a ₄₄	X4	

For the first run,	For the second run					
$\begin{bmatrix} 0.65\\ 1.0\\ 2.0\\ 2.85 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14}\\ a_{21} & a_{22} & a_{23} & a_{24}\\ a_{31} & a_{32} & a_{33} & a_{34}\\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \begin{bmatrix} 1\\ 0\\ 0\\ 0 \end{bmatrix}$	$\begin{bmatrix} 0.60\\ 1.0\\ 1.75\\ 2.5 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14}\\ a_{21} & a_{22} & a_{23} & a_{24}\\ a_{31} & a_{32} & a_{33} & a_{34}\\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \begin{bmatrix} 0\\ 1\\ 0\\ 0 \end{bmatrix}$					
$\mathbf{a_{11}} = 0.65, \mathbf{a_{21}} = 1.0, \mathbf{a_{31}} = 2.0, \mathbf{a_{41}} = 2.85$	$a_{12} = 0.60, a_{22} = 1.0, a_{32} = 1.75, a_{42} = 2.5$					

For the third run	For the fourth run
$[0.55]$ $[a_{11} \ a_{12} \ a_{13} \ a_{14}][0]$	$\begin{bmatrix} 0.70 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \end{bmatrix} \begin{bmatrix} 0 \end{bmatrix}$
1.0 a_{21} a_{22} a_{23} a_{24} 0	1.0 a_{21} a_{22} a_{23} a_{24} 0
$1.55 = a_{31} a_{32} a_{33} a_{34} 1$	$\begin{vmatrix} 2.3 \end{vmatrix} = \begin{vmatrix} a_{31} & a_{32} & a_{33} & a_{34} \end{vmatrix} \begin{vmatrix} 0 \end{vmatrix}$
$[2.2]$ $[a_{41} \ a_{42} \ a_{43} \ a_{44}][0]$	$[3.25]$ $[a_{41} \ a_{42} \ a_{43} \ a_{44}]$ $[1]$
$a_{13} = 0.55, a_{23} = 1.0, a_{33} = 1.55, a_{43}$	$a_{14} = 0.70, a_{24} = 1.0, a_{34} = 2.3, a_{44} = 3.25$

We will have [A] matrix when the values of the constants are substituted.

$$\begin{bmatrix} 0.65 & 0.60 & 0.55 & 0.70 \\ 1.0 & 1.0 & 1.0 & 1.0 \\ 2.0 & 1.75 & 1.55 & 2.3 \\ 2.85 & 2.5 & 2.2 & 3.25 \end{bmatrix}$$
For A₁₂;
$$\begin{bmatrix} Z1 \\ Z2 \\ Z3 \\ Z4 \end{bmatrix} = \begin{bmatrix} 0.65 & 0.60 & 0.55 & 0.70 \\ 1.0 & 1.0 & 1.0 & 1.0 \\ 2.0 & 1.75 & 1.55 & 2.3 \\ 2.85 & 2.5 & 2.2 & 3.25 \end{bmatrix} * \begin{bmatrix} 0.5 \\ 0.5 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} Z1 \\ Z2 \\ Z3 \\ Z4 \end{bmatrix} = \begin{bmatrix} 0.65 & 0.60 & 0.55 & 0.70 \\ 1.0 & 1.0 & 1.0 & 1.0 \\ 2.0 & 1.75 & 1.55 & 2.3 \\ 2.85 & 2.5 & 2.2 & 3.25 \end{bmatrix} * \begin{bmatrix} 0.5 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} Z1 \\ Z3 \\ Z4 \end{bmatrix} = \begin{bmatrix} 0.65 & 0.60 & 0.55 & 0.70 \\ 1.0 & 1.0 & 1.0 & 1.0 \\ 2.0 & 1.75 & 1.55 & 2.3 \\ 2.85 & 2.5 & 2.2 & 3.25 \end{bmatrix} * \begin{bmatrix} 0.5 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} Z1 \\ Z3 \\ Z4 \end{bmatrix} = \begin{bmatrix} 0.65 & 0.60 & 0.55 & 0.70 \\ 1.0 & 1.0 & 1.0 & 1.0 \\ 2.0 & 1.75 & 1.55 & 2.3 \\ 2.85 & 2.5 & 2.2 & 3.25 \end{bmatrix} * \begin{bmatrix} 0.5 \\ 0 \\ 0.5 \\ 0 \end{bmatrix}$$

$$Z1 = 0.65^{*}0.5 + 0.55^{*}0.5 = 0.60$$

$$Z2 = 1.0^{*}0.5 + 1.0^{*}0.5 = 1.0$$

Z3 = 2.0*0.5 + 1.55*0.5 = 1.775

For	A ₁₄	L					
[Z1]		0.65	0.60	0.55	0.70		[0.5]
Z2	_		1.0			÷	0
Z3	-	2.0	1.75	1.55	2.3	Ť	0
Z4			2.5				L0.5J
Z1 =	= 0.6	55*0.5	+0.70*	0.5 = 0	.675		
Z2 =	= 1.0	0*0.5 +	- 1.0*0.	5 = 1.0			
Z3 =	= 2.0)*0.5 +	-2.3*0.5	5 = 2.13	5		
Z4 =	= 2.8	85*0.5	+ 3.25	*0.5 = 3	3.05		

For	A_{24}	Ļ						
[Z1]		[0.65	0.60	0.55	0.70]		[0]	
Z2	_	1.0	1.0	1.0	1.0		0.5	
Z3	-	2.0	1.75	1.55	2.3	Ŧ	0	
Z4		2.85	0.60 1.0 1.75 2.5	2.2	3.25		0.5	

Z1 = 0.60*0.5 + 0.70*0.5 = 0.65Z2 = 1.0*0.5 + 1.0*0.5 = 1.0Z3 = 1.75*0.5 + 2.3*0.5 = 2.025Z4 = 2.5*0.5 + 3.25*0.5 = 2.875

]	For	A_{11}	12						
	[Z1]		0.65	0.60	0.55	0.70		[0.75]	
	Z2	_	1.0	1.0	1.0			0.25	
	Z3	-	2.0	1.75	1.55	2.3	Ŷ	0	
	Z4		2.85	2.5	2.2	3.25		$\begin{bmatrix} 0 \end{bmatrix}$	
2	Z1 =	0.0	55*0.7	5 + 0.60	*0.25 =	= 0.637	'5		
2	Z2 =	: 1.0)*0.75	+ 1.0*0	0.25 = 1	1.0			
2	Z3 =	2.0)*0.75	+1.75*	0.25 =	1.9375	5		
2	Z4 =	2.8	85*0.7	5 + 2.5	*0.25 =	2.762	5		

For	A ₁₁	14						
[Z1]		0.65	0.60	0.55	0.70		[0.75]	
Z2	_		1.0			÷	0	
Z3	-	2.0	1.75	1.55	2.3	Ŷ	0	
Z4		2.85	2.5	2.2	3.25		0.25	
Z1 =	= 0.6	55*0.7	5 + 0.70	*0.25 =	= 0.662	25		
Z2 =	: 1.(0*0.75	+ 1.0*0	0.25 = 1	1.0			
Z3 =	= 2.0)*0.75	+2.3*0	.25 = 2	.075			
Z4 =	= 2.8	35*0.7	5 + 3.25	5*0.25	= 2.95			

			For A	3334					
	[Z1]		0.65	0.60	0.55	0.70		[0]	
	Z2	_	1.0	1.0	1.0	1.0	*	0	
	Z3	-		1.75			Ŷ	0.75	
	Z4			2.5				0.25	
2	Z1 =	0.5	55*0.7	5 + 0.70	*0.25 =	= 0.587	5		
ļ	Z2 =	= 1.0)*0.75	+ 1.0*0	0.25 = 1	1.0			
ļ	Z3 =	= 1.5	55*0.7	5 +2.3*	0.25 =	1.7375	5		
ļ	Z4 =	= 2.2	2*0.75	+ 3.25*	*0.25 =	2.462	5		

For A₃₃₃₁

www.ijeijournal.comPage | 75

Z4 = 2.85*0.5 + 2.2*0.5 = 2.525

]	For	A ₂₃									
	[Z1]		0.65	0.60	0.55	0.70		[0]			
	Z2		1.0	1.0	1.0	1.0		0.5			
	Z3 Z4	=	2.0	1.75 2.5	1.55	2.3	*	0.5			
	Z4		2.85	2.5	2.2	3.25		0			
	Z1 = 0.60 * 0.5 + 0.55 * 0.5 = 0.575										
2	Z2 =	= 1.0)*0.5 +	- 1.0*0.	5 = 1.0						
1	Z3 =	= 1.7	75*0.5	+1.55*	0.5 = 1	.65					
1	Z4 =	= 2.	5*0.5	+ 2.2*0	.5 = 2.3	35					

		Fo	r A ₃₄					
[Z1]		0.65		0.55			[0]	
Z2	_	1.0	1.0	1.0	1.0	*	0	
Z3	-	2.0	1.75	1.55 2.2	2.3	^	0.5	
Z4							0.5	
Z1 =	0.5	55*0.5	+0.70*	0.5 = 0	.625			
Z2 =	: 1.0	0*0.5 +	1.0*0.	5 = 1.0				
Z3 =	: 1.5	55*0.5	+2.3*0	.5 = 1.9	925			
Z4 =	2.2	2*0.5 +	3.25*(0.5 = 2.7	725			

		Fo	r A ₁₁₁₃						
[Z	1]		0.65	0.60	0.55	0.70		[0.75]	
Z	2	_	1.0	1.0	1.0	1.0	*	0	
Z	3	-	2.0	1.75 2.5	1.55	2.3	*	0.25	
LZ4	4		2.85	2.5	2.2	3.25		0	
Z1	=	0.0	55*0.7	5 + 0.55	*0.25 =	= 0.625	5		
Z2	=	1.0	0*0.75	+ 1.0*0	0.25 = 1	1.0			
Z3	=	2.0	0*0.75	+1.55*	0.25 =	1.8875	5		
Z4	=	2.8	85*0.7	5 + 2.2	*0.25 =	2.687	5		

For	A ₂₂	23						
[Z1]		0.65	0.60	0.55	0.70		[0]	
Z2	_			1.0			0.75	
Z3	-	2.0	1.75	1.55	2.3	*	0.25	
Z4.		2.85	2.5	2.2	3.25		$\begin{bmatrix} 0 \end{bmatrix}$	
Z1 =	= 0.6	50*0.7	5 + 0.55	*0.25 =	= 0.587	'5		
Z2 =	= 1.0)*0.75	+ 1.0*0	0.25 = 1	1.0			
Z3 =	= 1.7	75*0.7	5+1.55	*0.25 =	= 1.70			
Z4 =	= 2.5	5*0.75	+ 2.2*0	0.25 = 2	2.425			

			For A ₁	222					
	[Z1]		0.65	0.60	0.55	0.70		[0.25]	
	Z2	_	1.0	1.0	1.0	1.0		0.75	
	Z3	-		1.75	1.55	2.3	*	0	
	Z4			2.5				0	
ļ	Z1 =	= 0.6	55*0.2	5 + 0.60	*0.75 =	=0.612	5		
ļ	Z2 =	= 1.0)*0.25	+ 1.0*0	0.75 = 1	1.0			
ļ	Z3 =	= 2.0	0*0.25	+1.75*	0.75 =	1.8125	5		
ļ	Z4 =	= 2.8	85*0.2	$5 + 2.5^{3}$	*0.75 =	2.587	5		

Z1 Z2 Z3 Z4	=	0.65 1.0 2.0	0.60 1.0 1.75 2.5	0.55 1.0 1.55 2 2	0.70 ⁻ 1.0 2.3 3.25	*	0.25 0 0.75
Z2 =	= 1.0	55*0.2 0*0.25	5 +0.55 + 1.0*(*0.75 = 0.75 = 1	= 0.575 1.0		
		0.20	+1.55* 5 + 2.2 ³ 1444	0.70	1.0020		
For	A ₁₂		1444				

		14							
	[Z1]		0.65	0.60	0.55	0.70		[0.25]	
	Z2	_	1.0	1.0	1.0	1.0	*	0.25	
	Z3	_		1.75			Ť	0.5	
	Z4		2.85	2.5	2.2	3.25			
2	Z1 =	: 0.6	55*0.2	5 + 0.60	*0.25 +	0.55*	0.5	5 = 0.5	875
2	Z2 =	: 1.0)*0.25	+ 1.0*0	0.25 + 1	1.0*0.5	i =	1.0	
	Z3 =	= 2.	0*0.25	5 +1.75 ³	*0.25 +	1.55*	0.5	= 1.7	125
2	Z4 =	2.8	35*0.2	$5 + 2.5^{\circ}$	*0.25 +	2.2*0.	5 =	2.437	5
For A ₁₃₄₄									

Fo	r A	\ ₁₁	23						
ΓZ	1]		0.65	0.60	0.55	0.70		[0.5]	
Z	2	_		1.0			÷	0.25	
Z	3	-	2.0	1.75	1.55	2.3	Ť	0.25	
LZ-	4			2.5				0	
Z1	=	0.6	55*0.5	+0.60*	0.25 +	0.55 +	0.2	5 = 0.0	5125
Z2	=	1.()*0.5 +	- 1.0*0.	25 + 1.	0*0.25	i =	1.0	
Z3	=	2.0)*0.5 +	-1.75*0	.25 + 1	.55*0.2	25 :	= 1.82	5
Z4	=	2.8	35*0.5	+ 2.5*().25 + 2	2.2*0.2	25 =	= 2.60	

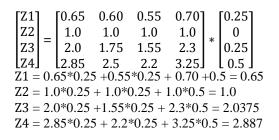
For A2344 [Z1] [0.65] 0.60 0.55 0.70] 0 Z2 1.0 1.0 1.0 1.0 0.25 = Z3 2.0 1.75 2.3 0.25 1.55 Z4 L2.85 2.5 2.2 3.25 0.5 Z1 = 0.60*0.25 + 0.55*0.25 + 0.70 + 0.5 = 0.6375Z2 = 1.0*0.25 + 1.0*0.25 + 1.0*0.5 = 1.0Z3 = 1.75*0.25 + 1.55*0.25 + 2.3*0.5 = 1.975Z4 = 2.5*0.25 + 2.2*0.25 + 3.25*0.5 = 2.80

For A₁₁₂₄ 0.70] 0.65 0.60 0.55 [0.5][Z1] Z2 1.0 1.0 0.25 1.0 1.0 * = Z3 2.0 1.75 1.55 2.3 0 LZ4 L2.85 2.5 2.2 3.25 L0.25 Z1 = 0.65*0.5 + 0.60*0.25 + 0.70 + 0.25 = 0.65Z2 = 1.0*0.5 + 1.0*0.25 + 1.0*0.25 = 1.0Z3 = 2.0*0.5 +1.75*0.25 + 2.3*0.25 = 2.0125 Z4 = 2.85*0.5 + 2.5*0.25 + 3.25*0.25 = 2.8625

For A₂₂₂₄

	44	24						
[Z1]		0.65	0.60	0.55	0.70		[0]	
Z2		1.0	1.0	1.0	1.0	÷	0.75	
Z3	-	2.0	0.60 1.0 1.75 2.5	1.55	2.3	Ŷ	0	
Z4.		2.85	2.5	2.2	3.25		0.25	
Z1 =	= 0.6	50*0.7	5 + 0.70	+0.25	= 0.62	5		
Z2 =	= 1.0	0*0.75	+ 1.0*0	0.25 = 1	1.0			

[0.65 0.60 0.55 [Z1] 0.701 [0.25] Z2 1.0 1.0 1.0 1.0 0 = Z3 2.0 1.75 1.55 2.3 0 Z4 L2.85 2.2 3.25 2.5 L0.75 $Z1 = 0.65 \times 0.25 + 0.70 \times 0.75 = 0.6875$ Z2 = 1.0*0.25 + 1.0*0.75 = 1.0Z3 = 2.0*0.25 +2.3*0.75 = 2.225 Z4 = 2.85*0.25 + 3.25*0.75 = 3.15



For	A ₁₁	23						
[Z1]		0.65	0.60	0.55	0.70		[0.5]	
Z2	_	1.0	1.0	1.0	1.0		0	
Z3	-	2.0	1.0 1.75	1.55	2.3	*	0.25	
Z4		2.85	2.5	2.2	3.25		0.25	
Z1 =	= 0.6		+0.55*				5 = 0.	6375
Z2 =	= 1.0	0*0.5 +	- 1.0*0.	25 + 1.	0*0.25	i =	1.0	
Z3 =	= 2.0	0*0.5 +	-1.55*0	0.25 + 2	.3*0.2	5 =	1.962	5
Z4 =	= 2.8	85*0.5	+ 2.2*0	0.25 + 3	3.25*0.	.25	= 2.78	875

For	A ₂₂	34						
[Z1]		[0.65	0.60	0.55	0.70		[0]	
Z2		1.0	1.0	1.0	1.0		0.5	
Z3	=	2.0	1.75			*	0.25	
LZ4.		2.85	2.5	2.2	3.25		0.25	
Z1 =	= 0.6	50*0.5	+0.55*	0.25 +	0.70 + 0.70	0.2	5 = 0.	6125
Z2 =	= 1.0	0*0.5 +	- 1.0*0.	25 + 1.	0*0.25	i =	1.0	
Z3 =	= 1.7	75*0.5	+1.55*	0.25 +	2.3*0.2	25 :	= 1.83	75
Z4 =	= 2.5	5*0.5 +	- 2.2*0.	25 + 3.	25*0.2	25 =	= 2.612	25

For A ₁₂	44				
[Z1]	[0.65	0.60	0.55	0.70]	[0.25]
Z2 _	1.0	1.0	1.0	1.0	0.25
Z3 =	2.0	1.75	1.55	2.3	
Z4	2.85	2.5	2.2	3.25	L 0.5]
Z1 = 0.6	55*0.2	5 + 0.60	*0.25 +	- 0.70 +0	0.5 = 0.6625
Z2 = 1.0)*0.25	+ 1.0*0	0.25 + 1	.0*0.5 =	= 1.0
Z3 = 2.0)*0.25	+1.75*	0.25 +	2.3*0.5	= 2.0875
Z4 = 2.8	85*0.2	5 + 2.5*	*0.25 +	3.25*0.	5 = 2.9625
Z3 = 1.7	75*0.75	5 + 2.3*	*0.25 =	1.8875	
Z4 = 2.5	5*0.75	+ 3.25*	*0.25 =	2.6875	
For A ₂₃	33				
[Z1]	0.65	0.60	0.55	0.70]	ן ס ן

1.0

1.55

2.2

Z2

LZ4

 $\begin{vmatrix} z_2 \\ z_3 \end{vmatrix} =$

1.0

2.0

L2.85

1.0

1.75

2.5

0.25

0.75

*

1.0

2.3

3.25

 $\begin{array}{l} Z1 = 0.60*0.25 + 0.55 + 0.75 = 0.5625 \\ Z2 = 1.0*0.25 + 1.0*0.75 = 1.0 \end{array}$

[Z1]			0.60				[0]	
Z2	_	1.0	1.0 1.75	1.0	1.0	÷	0.25	
Z3	-	2.0	1.75	1.55	2.3	Ŷ	0	
Z4		2.85	2.5	2.2	3.25		0.75	
Z1 =	= 0.6	50*0.2	5 + 0.70	+0.75	= 0.67	5		
Z2 =	= 1.0	0*0.25	+ 1.0*0	0.75 = 1	1.0			
Z3 =	= 1.7	75*0.2	5 + 2.3	*0.75 =	2.162	5		
Z4 =	= 2.4	5*0.25	+ 3.25*	*0.75 =	3.062	5		

Fe	or	A ₂₃	34						
٢Z	[1]		0.65	0.60	0.55	0.70		[0]	
Z	22	_	1.0	1.0	1.0	1.0	÷	0.25	
Z	23	-	2.0	1.75	1.55	2.3	Ŷ	0.5	
Lz	4		2.85	110	2.2	01201		0.25	
Z1	L =	0.0	50*0.2	5 + 0.55	*0.5+0).70 +0).25	5 = 0.6	60
Z_2	2 =	: 1.0	0*0.25	+ 1.0*0	0.5 + 1.	0*0.25	i =	1.0	
ZЗ	3 =	1.7	75*0.2	5 +1.55	*0.5 +	2.3*0.	25 :	= 1.78	75
Z4	1 =	2.5	5*0.25	+ 2.2*0	0.5 + 3.	25*0.2	25 =	= 2.53′	75

Z3 = 1.75*0.25 + 1.55*0.75 = 1.60Z4 = 2.5*0.25 + 2.2*0.75 = 2.27

For	A ₃₄	44						
[Z1]		0.65	0.60	0.55	0.70		[0]	
Z2		1.0	1.0	1.0	1.0		0	
Z3	=	2.0	1.75 2.5	1.55	2.3	*	0.25	
Z4		2.85	2.5	2.2	3.25		0.75	
Z1 =	0.5	55*0.2	5 + 0.70	+0.75 =	= 0.662	25		
Z2 =	= 1.0	0*0.25	+ 1.0*0	0.75 = 1	1.0			
Z3 =	= 1.5	55*0.2	5 + 2.3*	[∗] 0.75 =	2.112	5		
Z4 =	= 2.2	2*0.25	+ 3.25*	*0.75 =	2.937	5		

]	For	A_{12}	23						
	[Z1]		0.65	0.60	0.55	0.70		[0.25]	
	Z2	_	1.0	1.0	1.0	1.0	*	0.5	
	Z3	-	2.0	1.75	1.55	2.3	*	0.25	
	Z4		2.85	2.5	2.2	3.25		$\begin{bmatrix} 0 \end{bmatrix}$	
2	Z1 =	: 0.6	55*0.2	5 + 0.60	*0.5 +	0.55 + 0.55	0.2	5 = 0.6	50
2	Z2 =	: 1.()*0.25	+ 1.0*0	0.5 + 1.	0*0.25	=	1.0	
2	Z3 =	= 2.0)*0.25	+1.75*	0.5 + 1	.55*0.2	25 :	= 1.76	25
2	Z4 =	2.8	35*0.2	5 + 2.5*	*0.5 + 2	2.2*0.2	= 5	= 2.445	5

Z3 = 2.0*0.25 +1.55*0.5 + 2.3*0.25 = 1.85 Z4 = 2.85*0.25 + 2.2*0.5 + 3.25*0.25 = 2.625For A₁₂₂₄ 0.65 0.60 0.55 0.70] [0.25] [Z1] Z2 1.0 0.5 1.0 1.0 1.0 Z3 2.0 1.75 1.55 2.3 0 LZ4 L2.85 2.5 2.2 3.25 L0.25 Z1 = 0.65 * 0.25 + 0.60 * 0.5 + 0.70 + 0.25 = 0.6375Z2 = 1.0*0.25 + 1.0*0.5 + 1.0*0.25 = 1.0Z3 = 2.0*0.25 +1.75*0.5 + 2.3*0.25 = 1.95 Z4 = 2.85*0.25 + 2.5*0.5 + 3.25*0.25 = 2.775

For A₁₃₃₄ 0.65 0.60 0.55 0.70] [Z1] [0.25] Z21.0 1.0 1.0 1.0 0 * Z3 1.75 2.3 0.5 2.0 1.55 3.25 LZ4 2.85 2.5 2.2 0.25 $Z1 = 0.65 \times 0.25 + 0.55 \times 0.5 + 0.70 + 0.25 = 0.6063$ Z2 = 1.0*0.25 + 1.0*0.5 + 1.0*0.25 = 1.0

For A₁₂₃₄ 0.65 0.60 0.55 0.701 [0.25] [Z1] Z2 1.0 1.0 1.0 1.0 0.25 * = Z31.75 2.3 0.25 2.0 1.55 Z4J 3.25 2.85 2.5 2.2 L0.25 $Z1 = 0.65 \times 0.25 + 0.60 \times 0.25 + 0.55 \times 0.25 + 0.70 + 0.25 = 0.625$ Z2 = 1.0*0.25 + 1.0*0.25 + 1.0*0.5 + 1.0*0.25 = 1.0

 $Z_{2} = 1.0 \cdot 0.25 + 1.0 \cdot 0.25 + 1.0 \cdot 0.25 + 1.0 \cdot 0.25 = 1.0$ $Z_{3} = 2.0 \cdot 0.25 + 1.75 \cdot 0.25 + 1.55 \cdot 0.25 + 2.3 \cdot 0.25 = 1.90$ $Z_{4} = 2.85 \cdot 0.25 + 2.5 \cdot 0.25 + 2.2 \cdot 0.25 + 3.25 \cdot 0.25 = 2.70$

MIXTURE PROPORTIONS FOR CONTROL POINTS SHOWING ACTUAL AND PSEUDO COMPONENTS

Control points for A₁

[Z1] [0.65 0.60 0.55 0.70] [0.2] Z2 1.0 1.0 1.0 1.0 0.4 = 0.4 Z3 2.0 1.75 1.55 2.3 [Z4] L2.85 2.5 2.2 3.25 Z1 = 0.65*0.2 + 0.60*0.4 + 0.55*0.4 = 0.590 $\begin{array}{l} Z2=1.0*0.2+1.0*0.4+1.0*0.4=1.0\\ Z3=2.0*0.2+1.75*0.4+1.55*0.4=1.720\\ Z4=2.85*0.2+2.5*0.4+2.2*0.4=2.450\\ \textbf{Control points for } A_2 \end{array}$

Z2 = 1.0*0.2+1.0*0.4+1.0*0.4 = 1.0

Z3 = 2.0*0.2+1.75*0.4+2.3*0.4 = 2.020

Z4 = 2.85*0.2+2.5*0.4+3.25*0.4 = 2.870

[0.65 0.60 0.55 0.701[0.2] [Z1] Z2 1.0 1.0 1.0 1.0 0.4 = Z3 0 2.0 1.75 1.55 2.3 LZ4 L2.85 2.5 2.2 3.25 0.4 Z1 = 0.65*0.2 +0.60*0.4+0.70*0.4 = 0.650 **Control points A₃** [0.65] 0.60 0.55 0.701[0.2] [Z1] Z21.0 1.0 0 1.0 1.0 = Z3 1.75 2.0 1.55 2.3 0.4 LZ4J L2.85 2.2 3.25 0.4 2.5 Z1 = 0.65*0.2+0.55*0.4 +0.70*0.4 = 0.634 Z2 = 1.0*0.2+1.0*0.4+1.0*0.4 = 1.0Z3 = 2.0*0.2+1.55*0.4 +2.3*0.4 = 1.940 Z4 = 2.85*0.2+2.2*0.4 +3.25*0.4 = 2.750 **Control points A₄ Control points A5** 0.60 0.55 0.70] 0 0.65 [Z1] 0.4 Z21.0 1.0 1.0 1.0 Z3 2.0 1.75 1.55 2.3 0.4 Z4 3.25 L2.85 2.5 2.2 [0.2] Z1 = 0.60*0.4 + 0.55*0.4 + 0.70*0.2 = 0.600Z2 = 1.0*0.4+1.0*0.4+1.0*0.2 = 1.0Z3 = 1.75*0.4+1.55*0.4+2.3*0.2 = 1.78 Z4 = 2.5*0.4 + 2.2*0.4 + 3.25*0.2 = 2.530**Control points A7** 0.65 0.60 0.55 0.701 [0.4] Z1 Z21.0 1.0 1.0 1.0 0.4 = Z3 2.0 1.75 1.55 2.3 0 Z4 0.2 L2.85 2.5 2.2 3.25 Z1 = 0.65*0.4+0.60*0.4 +0.70*0.2= 0.640 Z2 = 1.0*0.4 + 1.0*0.4 + 1.0*0.2 = 1.0Z3 = 2.0*0.4 + 1.75*0.4 + 2.3*0.2 = 1.960Z4 = 2.85*0.4 + 2.5*0.4 + 3.25*0.2 = 2.790**Control points A**₉ [0.65 0.60 0.55 0.70] **[**0.4] [Z1] Z20.4 1.0 1.0 1.0 1.0 = Z3 2.0 1.75 1.55 2.3 0.2 Z4 2.85 2.5 2.2 3.25 Z1 = 0.65 * 0.4 + 0.60 * 0.4 + 0.55 * 0.2 = 0.610Z2 = 1.0*0.4+1.0*0.4+1.0*0.2 = 1.0Z3 = 2.0*0.4+1.75*0.4+ 1.55*0.2 = 1.810 Z4 = 2.85*0.4 + 2.5*0.4 + 2.2*0.2 = 2.580Control pointsA₁₀ Control points A₁₁ 0.65 0.60 0.55 0.701 0 [Z1] Z2 1.0 0.35 1.0 1.0 1.0 Z3 2.0 1.75 1.55 2.3 0.35 [Z4] 2.85 3.25. 2.5 2.2 L 0.3 . Z1 = 0.60*0.35+0.55*0.35 +0.70*0.3 = 0.6125 Z2 = 1.0*0.35+1.0*0.35+1.0*0.3 = 1.0Z3 = 1.75*0.35+1.55*0.35+2.3*0.3 = 1.8450Z4 = 2.5*0.35+2.2*0.35 +3.25*0.3 = 2.620 Control points A₁₃ [Z1] 0.65 0.60 0.55 0.70][0.35] Z2 1.0 1.0 1.0 1.0 0.35 Z3 2.0 1.75 1.55 2.3 0 LZ4J L2.85 2.2 3.25 L 0.3 2.5 Z1 = 0.65*0.35+0.60*0.35 +0.70*0.3= 0.6475 Z2 = 1.0*0.35+1.0*0.35 +1.0*0.3= 1.0

0.55 [Z1] [0.65] 0.60 0.70] 0 . Z2 1.0 1.0 1.0 1.0 0.2 = Z3 2.0 1.55 2.3 0.4 1.75 LZ4 L2.85 2.5 2.2 3.25 0.4 $Z1 = 0.60 \times 0.2 + 0.55 \times 0.4 + 0.70 \times 0.4 = 0.620$ Z2 = 1.0*0.2+1.0*0.4+1.0*0.4 = 1.0Z3 = 1.75*0.2+1.55*0.4 + 2.3*0.4 = 1.890Z4 = 2.5*0.2+2.2*0.4 +3.25*0.4 = 2.680 Control points A₆ 0.55 [0.4] 0.60 0.70] 0.65 [Z1] Z2 1.0 1.0 0 1.0 1.0 Z3 2.0 1.75 1.55 2.3 0.4 Z4 L2.852.5 2.2 3.25 [0.2] Z1 = 0.65*0.4+0.55*0.4 +0.70*0.2= 0.620 Z2 = 1.0*0.4+1.0*0.4+1.0*0.2 = 1.0Z3 = 2.0*0.4+1.55*0.4+2.3*0.2 = 1.880 Z4 = 2.85*0.4+2.2*0.4 +3.25*0.2 = 2.670 **Control points A₈** 0.55 Z1 0.65 0.60 0.701 [0.4] Ζ2 1.0 1.0 1.0 1.0 0.4 = Z3 2.0 1.75 1.55 2.3 0.2 Z4 2 85 2.5 2.2 3.25 10 Z1 = 0.65*0.4+0.60*0.4 +0.55*0.2= 0.610 Z2 = 1.0*0.4 + 1.0*0.4 + 1.0*0.2 = 1.0Z3 = 2.0*0.4 + 1.75*0.4 + 1.55*0.2 = 1.810Z4 = 2.85*0.4+2.5*0.4 +2.2*0.2 = 2.580 0.65 0.60 0.55 0.70] [0.25] FZ11 Z2 0.25 1.0 1.0 1.0 1.0 = Z3 2.0 1.75 1.55 2.3 0.25 LZ4 L2.852.5 2.2 3.25 L0.25. $Z1 = 0.65 \times 0.25 + 0.60 \times 0.25 + 0.55 \times 0.25 + 0.70 \times 0.25 =$ 0.6250 Z2 = 1.0*0.25+1.0*0.25+1.0*0.25+1.0*0.25 = 1.0Z3 = 2.0*0.25+1.75*0.25+1.55*0.25+2.3*0.25 =1.90 Z4 = 2.85*0.25+2.5*0.25 +2.280.25 +3.25*0.25 = 2.70Control points A₁₂ 0.65 0.60 0.55 0.701 0.3 [Z1] Z2 1.0 0 1.0 1.0 1.0 Z3 2.0 1.75 1.55 2.3 0.35 Z4 2.85 2.2 3.25. 0.35 2.5 Z1 = 0.65*0.3+0.55*0.35 +0.70*0.35= 0.6325 Z2 = 1.0*0.3+1.0*0.35 +1.0*0.35= 1.0 Z3 = 2.0*0.3+1.55*0.35+ 2.3*0.35 = 1.9475 Z4 = 2.85*0.3+2.2*0.35+3.25*0.35 = 2.7625 Z3 = 2.0*0.35 + 1.75*0.35 + 2.3*0.3 = 2.0025Z4 = 2.85*0.35+2.5*0.35+3.25*0.3 = 2.8475 Control points A₁₄ 0.65 0.60 0.55 0.70 0.3 [Z1] Z2 1.0 1.0 1.0 0.35 1.0

*

0.35

L 0 _

2.3

3.25

=

2.0

L2.85

1.75

2.5

1.55

2.2

Z3

LZ4

Z1 = 0.65*0.3+0.60*0.35 +0.55*0.35= 0.5975 Z2 = 1.0*0.3+1.0*0.35+1.0*0.35=1.0Control points A₁₅ 0.65 0.55 0.701 0 **[**71] 0.60 Z20.3 1.0 1.0 1.0 1.0 = Z32.0 1.75 1.55 2.3 0.35 Z4 3.25 2.85 2.5 2.2 L0.35 Z1 = 0.60*0.3+0.55*0.35 +0.70*0.35= 0.6175 Z2 = 1.0*0.3+1.0*0.35 +1.0*0.35= 1.0 Z3 = 1.75*0.3+1.55*0.35+2.3*0.35 = 1.8726Z4 = 2.5*0.3+2.2*0.35+3.25*0.35 = 2.6575

Z3 = 2.0*0.3+1.75*0.35+1.55*0.35 = 1.755Z4 = 2.85*0.3+2.5*0.35+2.2*0.35 = 2.50

MATERIALS

III. MATERIALS AND METHOD

Dangote3x Ordinary Portland cement product conforming to BS and ASTM standards with minimum rate of hardening, andnatural river sharp sand used as fine aggregate were obtained in kaura Namoda, Zamfara state, Nigeria; with maximum size fine aggregate being 4.75mm and treated to be free from impurities. Palm kernel shell used as 100% replacement for granite, was obtained from palm oil producing areas of Edo state, Nigeria. The crushed shells were dried and sieved to get rid of impurities. The water used for mixing and curing is potable drinking water from bored hole at the Federal Polytechnic Kaura Namoda, Zamfara state, and tested to be suitable for concrete work.

METHODS

Required quantity of the constituent materials were first thoroughly dry mixed in a manually operated concrete mixer before water is added for wet mixing and casting of cubes for the strength test.

The compressive strength test was performed in accordance with BS 1881- 116 and ACI 311.6 - 18 specifications using Magnus Compression Testing Machine and concrete cubes of sizes of (150x150x150) mm. A total of seventy (70) cubes were produced for the experimental process for the thirty-five (35) points, while fifteen points of control was considered and a total of thirty cubes were produced as control.

Two specimen of each mix was crushed at the curing regime of 28 days and the average recorded as the strength achieved.

The compressive strength was determined using the equation; $Cs = \frac{P}{A}$ where,

		PSEU	DO			ACTUAL			
POINTS	W/C	CEMENT	SAND	PKS	RESPONSE	W/C	CEMENT	SAND	PKS
	X1	X2	X3	X4	Y	Z1	Z2	Z3	Z4
1	1	0	0	0		0.65	1	2	2.85
	0	1	0	0		0.6	1	1.75	2.5
3	0	0	1	0		0.55	1	1.55	2.2
4	0	0	0	1		0.7	1	2.3	3.25
5	0.5	0.5	0	0		0.625	1	1.875	2.675
6	0.5	0	0.5	0		0.6	1	1.775	2.525
7	0.5	0	0	0.5		0.675	1	2.15	3.05
8	0	0.5	0.5	0		0.575	1	1.65	2.35
9	0	0.5	0	0.5		0.65	1	2.025	2.875
10	0	0	0.5	0.5		0.625	1	1.925	2.725
11	0.75	0.25	0	0		0.6375	1	1.9375	2.7625
12	0.75	0	0.25	0		0.625	1	1.8875	2.6875
13	0.75	0	0	0.25		0.6625	1	2.075	2.95
14	0	0.75	0.25	0		0.5875	1	1.7	2.425
15	0	0	0.75	0.25		0.5875	1	1.7375	2.4625

 Table 1: Matrix for Scheffe`s (4,4) Lattice Polynomial

Optimization of the	compressive stren	ngth of Palm Kernel	l Shell lightweight Aggregate con	crete
- I - S J J	T T T			

16	0.25	0.75	0	0	0.6125	1	1.8125	2.5875
17	0.25	0	0.75	0	0.575	1	1.6625	2.3625
18	0.25	0	0	0.75	0.6875	1	2.225	3.15
19	0.25	0.25	0.5	0	0.5875	1	1.7125	2.4375
20	0.25	0	0.25	0.5	0.65	1	2.0375	2.8875
21	0.5	0.25	0.25	0	0.6125	1	1.825	2.6
22	0.5	0	0.25	0.25	0.6375	1	1.9625	2.7875
23	0	0.25	0.25	0.5	0.6375	1	1.975	2.8
24	0	0.5	0.25	0.25	0.6125	1	1.8375	2.6125
25	0.5	0.25	0	0.25	0.65	1	2.0125	2.8625
26	0.25	0.25	0	0.5	0.6625	1	2.0875	2.9625
27	0	0.75	0	0.25	0.625	1	1.8875	2.6875
28	0	0.25	0.75	0	0.5625	1	1.6	2.275
29	0	0.25	0	0.75	0.675	1	2.1625	3.0625
30	0	0	0.25	0.75	0.6625	1	2.1125	2.9875
31	0	0.25	0.5	0.25	0.6	1	1.7875	2.5375
32	0.25	0.5	0.25	0	0.6	1	1.7625	2.5125
33	0.25	0	0.5	0.25	0.6063	1	1.85	2.625
34	0.25	0.5	0	0.25	0.6375	1	1.95	2.775
35	0.25	0.25	0.25	0.25	0.625	1	1.9	2.7

Table 2: Mixture Proportion of Control points showing pseudo and actual

Points		Pse	eudo				Ac	ctual	
	w/c	Cement	Sand	PKS	Response	w/c	Cement	Sand	PKS
	X1	X2	X3	X4	Yc	Z1	Z2	Z3	Z4
1	0.2	0.4	0.4	0	Yc ₁	0.59	1	1.72	2.45
2	0.2	0.4	0	0.4	Yc ₂	0.65	1	2.02	2.87
3	0.2	0	0.4	0.4	Yc ₃	0.63	1	1.94	2.75
4	0	0.2	0.4	0.4	Yc_4	0.62	1	1.89	2.68
5	0	0.4	0.4	0.2	Yc ₅	0.6	1	1.78	2.53
6	0.4	0	0.4	0.2	Yc ₆	0.62	1	1.88	2.67
7	0.4	0.4	0	0.2	Yc ₇	0.64	1	1.96	2.79
8	0.4	0.4	0.2	0	Yc ₈	0.61	1	1.81	2.58
9	0.4	0.4	0.2	0	Yc ₉	0.61	1	1.81	2.58
10	0.25	0.25	0.25	0.25	Yc ₁₀	0.625	1	1.9	2.7
11	0	0.35	0.35	0.3	Yc11	0.6125	1	1.845	2.62
12	0.3	0	0.35	0.35	Yc ₁₂	0.6325	1	1.9475	2.7625
13	0.35	0.35	0	0.3	Yc ₁₃	0.6475	1	2.0025	2.8475
14	0.3	0.35	0.35	0	Yc ₁₄	0.5975	1	1.755	2.5
15	0	0.3	0.35	0.35	Yc ₁₅	0.6175	1	1.8726	2.6575

IV. RESULTS AND DISCUSSION

The results of the compressive strength for palm kernel shells concrete were obtained from laboratory test on 150x150x150 concrete cube specimens after 28 days curing using compression machine. The results are presented in tables 3 and 4, for the control points.

	Table 5. Offaxial Compressive Strength of Concrete										
Sample	Curing	Failure	Load (KN)	Area (mm ²)	Compressi	Compressive Strength(Nmm ⁻²) MPa					
Points	Age	А	В		А	В	Average				
N1	28days	35	40	22500	1.556	1.778	1.667				
N ₂	28days	40	45	22500	1.778	2.000	1.889				
N ₃	28days	85	86	22500	3.778	3.822	3.800				
N_4	28days	35	40	22500	1.556	1.778	1.667				
N ₅	28days	60	45	22500	2.667	2.000	2.334				
N ₆	28days	65	50	22500	2.889	2.222	2.556				

Table 3: Uniaxial Compressive Strength of Concrete

Optimization of the compressive strength of Palm Kernel Shell lightweight Aggregate concrete ..

N_7	28days	60	60	22500	2.667	2.667	2.667
N_8	28days	65	70	22500	2.889	3.111	3.000
N_9	28days	45	50	22500	2.000	2.222	2.111
N ₁₀	28days	50	50	22500	2.222	2.222	2.222
N ₁₁	28days	60	65	22500	2.667	2.889	2.778
N ₁₂	28days	55	60	22500	2.444	2.667	2.556
N ₁₃	28days	55	50	22500	2.444	2.222	2.333
N_{14}	28days	60	65	22500	2.667	2.889	2.778
N ₁₅	28days	70	60	22500	3.111	2.667	2.889
N ₁₆	28days	70	75	22500	3.111	3.333	3.222
N ₁₇	28days	65	50	22500	2.889	2.222	2.556
N ₁₈	28days	55	50	22500	2.444	2.222	2.333
N ₁₉	28days	70	70	22500	3.111	3.111	3.111
N ₂₀	28days	60	55	22500	2.667	2.444	2.556
N ₂₁	28days	35	50	22500	1.556	2.222	1.889
N ₂₂	28days	45	55	22500	2.000	2.444	2.222
N ₂₃	28days	50	60	22500	2.222	2.667	2.445
N ₂₄	28days	70	65	22500	3.111	2.889	3.000
N ₂₅	28days	60	60	22500	2.667	2.667	2.667
N ₂₆	28days	55	60	22500	2.444	2.667	2.556
N ₂₇	28days	65	60	22500	2.889	2.667	2.778
N ₂₈	28days	60	65	22500	2.667	2.889	2.778
N ₂₉	28days	35	40	22500	1.556	1.778	1.667
N ₃₀	28days	55	60	22500	2.444	2.667	2.556
N ₃₁	28days	50	60	22500	2.222	2.667	2.445
N ₃₂	28days	65	70	22500	2.889	3.111	3.000
N ₃₃	28days	60	60	22500	2.667	2.667	2.667
N ₃₄	28days	65	70	22500	2.889	3.111	3.000
N ₃₅	28days	65	55	22500	2.889	2.444	2.667

 Table 4:
 (4, 4) Lattice Control Mix. (28 days compressive Strength of PKS Concrete)

Sample	Curing age	Failure Loads	(KN)	Area	Compressive	Strength (N/mn	n²) Mpa
Points		Α	В	(mm ²)	А	B Ave	erage
C1	28 (days)	38	40	22500	1.689	1.778	1.733
C_2	28	86	86	22500	3.822	3.822	3.822
C ₃	28	65	55	22500	2.889	2.444	2.667
C_4	28	60	60	22500	2.667	2.667	2.667
C ₅	28	45	50	22500	2.000	2.222	2.111
C_6	28	60	65	22500	2.667	2.889	2.778
C ₇	28	60	65	22500	2.667	2.889	2.778
C_8	28	65	50	22500	2.889	2.222	2.556
C ₉	28	70	70	22500	3.111	3.111	3.111
C ₁₀	28	46	55	22500	2.044	2.444	2.244
C ₁₁	28	70	65	22500	3.111	2.889	3.000
C ₁₂	28	50	60	22500	2.000	2.667	2.333
C ₁₃	28	38	40	22500	1.689	1.778	1.733
C ₁₄	28	60	60	22500	2.667	2.667	2.667
C ₁₅	28	65	57	22500	2.889	2.533	2.711

REGRESSION EQUATION FOR COMPRESSIVE STRENGTH

The coefficients of the fourth-degree polynomial are determined as presented below

$\alpha 1 = 1.667$	$\alpha 2 = 1.889$	a3 = 3.8	a4 = 1.667	α12 =2.222	$\alpha 13 = -1.156$	a14 = -0.266
$\alpha 23 = 0.622$	$\alpha 24 = -2.49$	a34 =-2.046	$\lambda 12 = -1.776$	$\lambda 13 = 5.688$	$\lambda 14 = 5.921E-16$	$\lambda 23 = 5.096$
$\lambda 24 = 5.925$	λ34 =23.349	$\mu 12 = 17.181$	µ13 = -0.949	µ14=-1.792	$\mu 23 = 1.781$	$\mu 24 = 3.561$
$\mu 34 = 7.944$	all23=-101.085	a1124=-17.176	a1134=33.395	a1223=17.419	$\alpha 1224 = 14.464$	$\alpha 1233 = 79.747$
a1234=5.872	a1244 = 11.835		$\alpha 1344 = 3.200$	$\alpha 2234 = 7.107$	$\alpha 2334 = -13.864$	a2344= 10.325
		40.096				

Substituting these values of coefficients into Scheffe's model equation will give;

 $\hat{y} = 1.667x_1 + 1.889x_2 + 3.8x_3 + 1.667x4 + 2.222x_1x_2 - 1.156x_1x_3 - 0.266x_1x_4$

 $+0.622 x_2 x_3-2.49 x_2 x_4-2.046 x_3 x_4-1.776 x_1 x_2 \left(x_1\text{-} x_2\right)+5.688 x_1 x_3 (x_1\text{-} x_3)$

 $+ 5.921E - 16x_1x_4 \left(x_1 - x_4\right) + 5.096x_2x_3(x_2 - x_3) + 5.925x_2x_4(x_2 - x_4) + 2\ 3.349x_3x_4 \left(x_3 - x_4\right) + 2 \ 3.349x_4 \left(x_3 - x_4\right) + 2 \ 3.349x_4 \left(x_4 - x_4\right) + 2$

 $\begin{array}{l} + 17.181x_{1}x_{2}\left(x_{1}-x_{2}\right)^{2} - 0.949x_{1}x_{3}(x_{1}-x_{3})^{2} - 1.792x_{1}x_{4}(x_{1}-x_{4})^{2} \\ + 1.781x_{2}x_{3}(x_{2}-x_{3})^{2} + 3.561x_{2}x_{4}\left(x_{2}-x_{4}\right)^{2} + 7.944x_{3}x_{4}(x_{3}-x_{4})^{2} - 101.085x_{1}^{2}x_{2}x_{3} \\ - 17.176x_{1}^{2}x_{2}x_{4} + 33.395x_{1}^{2}x_{3}x_{4} + 17.419x_{1}x_{2}^{2}x_{3} + 14.464x_{1}x_{2}^{2}x_{4} \\ + 7.107x_{2}^{2}x_{3}x_{4} + 40.096x_{1}x_{3}^{2}x_{4} - 13.864x_{2}x_{3}^{2}x_{4} + 79.747x_{1}x_{2}x_{3}^{2} \end{array}$

 $+ 11.835 x_1 x_2 {x_4}^2 + 3.20 x_1 {x_3} {x_4}^2 + 10.325 x_2 {x_3} {x_4}^2 + 5.872 x_1 {x_2} {x_3} {x_4} \\$

This is the improved model for the optimization of the compressive strength of palm kernel shell concrete using Scheffe's fourth degree polynomials.

REPLICATE VARIANCE

No. of Expt. Pts	Replicates	Response yi	Response Symbol	mi	n vana ỹ		S _i ²
(N)	Replicates	(N/mm ²)	Response by moor	∑yi i=1	J	m_i $\sum_{i=1}^{m_i} yi^2$	51
				1=1		i=1	
1	1P	1.556	y 1	3.334	1.667	5.5824	0.0246
	1Q	1.778					
2	2P	1.778	y ₂	3.778	1.889	7.1613	0.0246
	2Q	2					
3	3P	3.778	y ₃	7.6	3.8	28.8810	0.0010
	3Q	3.822					
4	4P	1.556	y 4	3.334	1.667	5.5824	0.0246
	4Q	1.778					
5	5P	2.667	y12	4.667	2.3335	11.1129	0.2224
	5Q	2					
6	6P	2.889	y 13	5.111	2.5555	13.2836	0.2224
	6Q	2.222					
7	7P	2.667	y 14	5.334	2.667	14.2258	0.0000
	7Q	2.667					
8	8P	2.889	y ₂₃	6	3	18.0246	0.0246
	8Q	3.111					
9	9P	2	y ₂₄	4.222	2.111	8.9373	0.0246
	9Q	2.222					
10	10P	2.222	y 34	4.444	2.222	9.8746	0.0000
	10Q	2.222					
11	11P	2.667	y 1112	5.556	2.778	15.4592	0.0246
	11Q	2.889					
12	12P	2.444	y 1113	5.111	2.5555	13.0860	0.0249
	12Q	2.667					
13	13P	2.444	y 1114	4.666	2.333	10.9104	0.0246
	13Q	2.222					
14	14P	2.667	y 2223	5.556	2.778	15.4592	0.0246
	14Q	2.889					
15	15P	3.111	y 3334	5.778	2.889	16.7912	0.0986
	15Q	2.667					
16	16P	3.111	y 1222	6.444	3.222	20.7872	0.0246
	16Q	3.333					
					1		

 Table 5: Compressive Strength Test Results and Replication Variance of Response

17	17P	2.889	y ₁₃₃₃	5.111	2.5555	13.2836	0.2224
	17Q	2.222					
18	18P	2.444	y 1444	4.666	2.333	10.9104	0.0246
	18Q	2.222					
19	19P	3.111	y ₁₂₃₃	6.222	3.111	19.3566	0.0000
	19Q	3.111					
20	20P	2.667	y 1344	5.111	2.5555	13.0860	0.0249
	20Q	2.444					
21	21P	1.556	y 1123	3.778	1.889	7.3584	0.2218
	21Q	2.222					
22	22P	2	y ₁₁₃₄	4.444	2.222	9.9731	0.0986
	22Q	2.444					
23	23P	2.222	y 2344	4.889	2.4445	12.0502	0.0990
	23Q	2.667					
24	24P	3.111	y ₂₂₃₄	6	3	18.0246	0.0246
	24Q	2.889					
25	25P	2.667	y 1124	5.334	2.667	14.2258	0.0000
	25Q	2.667					
26	26P	2.444	y 1244	5.111	2.5555	13.0860	0.0249
	26Q	2.667					
27	27P	2.889	y 2224	5.556	2.778	15.4592	0.0246
	27Q	2.667					
28	28P	2.667	y 2333	5.556	2.778	15.4592	0.0246
	28Q	2.889					
29	29P	1.556	y 2444	3.334	1.667	5.5824	0.0246
	29Q	1.778					
30	30P	2.444	y ₃₄₄₄	5.111	2.5555	13.0860	0.0249
	30Q	2.667					
31	31P	2.222	y 2334	4.889	2.4445	12.0502	0.0990
	31Q	2.667					
32	32P	2.889	y ₁₂₂₃	6	3	18.0246	0.0246
	32Q	3.111					
33	33P	2.667	y 1334	5.334	2.667	14.2258	0.0000
	33Q	2.667					
34	34P	2.889	y 1224	6	3	18.0246	0.0246
	34Q	3.111					
35	35P	2.889	y ₁₂₃₄	5.333	2.6665	14.3195	0.0990
	35Q	2.444					
36	36P	1.689	C1	3.467	1.7335	6.0140	0.0040
	36Q	1.778					
<u>.</u>			1	1	1	1	

37	37P	3.822	C2	7.644	3.822	29.2154	0.0000
	37Q	3.822					
38	38P	2.889	C3	5.333	2.6665	14.3195	0.0990
	38Q	2.444					
39	39P	2.667	C4	5.334	2.667	14.2258	0.0000
	39Q	2.667					
40	40P	2	C5	4.222	2.111	8.9373	0.0246
	40Q	2.222					
41	41P	2.667	C6	5.556	2.778	15.4592	0.0246
	41Q	2.889					
42	42P	2.667	C7	5.556	2.778	15.4592	0.0246
	42Q	2.889					
43	43P	2.889	C8	5.111	2.5555	13.2836	0.2224
	43Q	2.222					
44	44P	3.111	C9	6.222	3.111	19.3566	0.0000
	44Q	3.111					
45	45P	2.044	C10	4.488	2.244	10.1511	0.0800
	45Q	2.444					
46	46P	3.111	C11	6	3	18.0246	0.0246
	46Q	2.889					
47	47P	2	C12	4.667	2.3335	11.1129	0.2224
	47Q	2.667					
48	48P	1.689	C13	3.467	1.7335	6.0140	0.0040
	48Q	1.778					
49	49P	2.667	C14	5.334	2.667	14.2258	0.0000
	49Q	2.667	1				
50	50P	2.889	C15	5.422	2.711	14.7624	0.0634
	50Q	2.533	1				
		2.333	l.	I	1	2	
						□ Sf =	2.6717

The mean response Y and variance of replicates S_i^2 presented in table 5 above are obtained as follow;

$$Y = \frac{\sum_{i=1}^{n} Y_i}{\binom{n}{1}}$$

$$S_i^2 = \left| \frac{1}{n-1} \right| \left[\sum Y_i^2 - \left[\frac{1}{n(\sum Y_i)^2} \right] \right]$$

Therefore,
$$S_i^2 = \left| \frac{1}{n-1} \right| \left[\sum_{i=1}^n [Y_i - Y]^2 \right]$$
$$V_e = \left(\sum N \right) - 2 = 50 - 2 = 48$$
$$\therefore \text{ Replicate variance,} \qquad S_i^2 = \frac{2.6717}{48} = 0.0557$$

Replicate error $S_{i=\sqrt{0.0557}} = 0.236$

The experimental results and Scheffe`s model test results for the concrete compressive strength are presented as shown in table 6

Sample Points	Compressive strength $Y_{exp}(N/mm^2)$	Compressive strength Y _{Pred} (N/mm ²)
N_1	1.667	1.667
N_2	1.889	1.889
N_3	3.800	3.800
N_4	1.667	1.667
N_5	2.334	2.334
N_6	2.556	2.445
N_7	2.667	1.601
N ₈	3.000	3.000
N ₉	2.111	1.156
N ₁₀	2.222	2.222
N ₁₁	2.778	2.778
N ₁₂	2.556	2.472
N ₁₃	2.333	1.533
N ₁₄	2.778	3.045
N ₁₅	2.889	5.445
N ₁₅	3.222	3.222
N ₁₆	2.556	2.472
N ₁₇ N ₁₈	2.333	1.533
	3.111	3.123
N ₁₉		
N ₂₀	2.556	1.782
N ₂₁	1.889	1.857
N ₂₂	2.222	2.911
N ₂₃	2.445	1.011
N ₂₄	3.000	2.419
N ₂₅	2.667	1.813
N ₂₆	2.556	1.509
N ₂₇	2.778	2.089
N ₂₈	2.778	3.045
N ₂₉	1.667	0.867
N ₃₀	2.556	0.000
N ₃₁	2.445	3.021
N ₃₂	3.000	3.063
N ₃₃	2.667	3.836
N ₃₄	3.000	2.315
N ₃₅	2.667	2.418
C ₁	1.733	1.667
C_2	3.822	3.380
C ₃	2.667	1.818
C ₄	2.667	2.747
C ₅	2.111	1.929
C ₆	2.778	2.830
C ₇	2.778	3.505
C ₇	2.556	2.032
C ₈ C ₉	3.111	2.032
	2.244	
C ₁₀	3.000	2.110 2.418
C ₁₁		
C ₁₂	2.333	2.275
C ₁₃	1.733	2.908
C ₁₄	2.667	1.855
C ₁₅	2.711	2.798

Table 6: Experimental and Predicted values of 28 days compressive Strength of PKS Concrete

V. VALIDATION AND TEST OF ADEQUACY OF MODEL

Statistical analysis using student's t-test and ANOVA was employed to analyse the improved Scheffe's model where the adequacy of the model was tested against the experimental results of the control points. The predicted values (Y-predicted) for the test control points were determined after substituting the corresponding values of X_1 , X_2 , X_3 , and X_4 , into Scheffe's model equation. The predicted values and the experimental values (Y- pred. and Y-expt), were compared. To test for adequacy of the model, student's t-test and ANOVA were used at 95% confidence level on the compressive strength at the control points of C_1 , C_2 , C_3 , C_4 , C_5 , C_6 , C_7 , C_8 , C_9 , C_{10} , C_{11} , C_{12} , C_{13} , C_{14} , C_{15} .

Table 7 : Student t-test for 28 days compressive strength of concrete							
		Compressive Str	ength (N/mm ²)	Students t-test			
Points	Curing Age	Yexpt	ypred	$\mathbf{Y} = \mathbf{y}_{expt} - \mathbf{y}_{pred}$	Y _a -Y	$(\mathbf{Y}_{\mathbf{a}}\mathbf{-}\mathbf{Y})^2$	
C1	28 DAYS	1.733	3.380	-1.647	1.920	3.687	
C2	28 DAYS	3.822	1.818	2.004	-1.731	2.995	
C ₃	28 DAYS	2.667	2.747	-0.080	0.353	0.125	
C_4	28 DAYS	2.667	1.929	0.738	-0.465	0.216	
C ₅	28 DAYS	2.111	2.830	-0.719	0.993	0.985	
C ₆	28 DAYS	2.778	3.505	-0.727	1.000	1.000	
C ₇	28 DAYS	2.778	2.032	0.746	-0.472	0.223	
C ₈	28 DAYS	2.556	2.110	0.446	-0.173	0.030	
C ₉	28 DAYS	3.111	2.110	1.001	-0.728	0.530	
C ₁₀	28 DAYS	2.244	2.418	-0.174	0.447	0.200	
C ₁₁	28 DAYS	3	2.275	0.725	-0.452	0.204	
C ₁₂	28 DAYS	2.333	2.908	-0.575	0.848	0.720	
C ₁₃	28 DAYS	1.733	1.855	-0.122	0.395	0.156	
C ₁₄	28 DAYS	2.667	2.798	-0.131	0.405	0.164	
C ₁₅	28 DAYS	2.711	2.010	0.701	-0.428	0.183	
TOTAL				2.186		11.418	
AVERAGE (Y _{a)}				0.273			

Table 7 : Student t	-test for 28 days con	mpressive strength of concrete

$$\mathbf{t}_{\text{stat}} = \frac{\Sigma(\text{exp-pred})}{\sqrt{\frac{\left(15^*\Sigma((\text{exp-pred})^2) \cdot (\Sigma(\text{exp-pred}))\right)^2}{15 \cdot 1}}} = \frac{(2.186)}{\sqrt{\frac{(15^*11.418) \cdot (2.186^2)}{(14)}}} = \mathbf{0.634}$$

At 95% confidence level the significant level is 0.05. For the two-tailed t-test, the significant level $\alpha = 0.05$ and 0.025 and using table 7, t_{stat} is calculated. $t_{\text{critical}}\!=\!2.145$

 $t_{sat.} = 0.634$ and $t_{crit.} = 2.145$, implies that $t_{sat} < t_{crit.}$ but between - 2.145 and 2.145, indication of a good correlation

ANALYSIS OF VARIANCE

From the result in table 9, F = 0.7293 and $F_{crit} = 4.2582$ (F- distribution table). This means that $F_{crit.}$ >F hence, there is no significant difference between the experimental and the model results. The model is therefore adequate to use in predicting the split tensile strength when the mix ratio is known and vice versa.

	Table 8: The Compressive Strength's ANOVA					
Point	(Expt.)	(Pred.)	(Expt.) ²	$(\mathbf{Pred.})^2$		
C1	1.733	3.380	3.003	11.424		
C2	3.822	1.818	14.608	3.305		
C3	2.667	2.747	7.113	7.546		
C4	2.667	1.929	7.113	3.721		
C5	2.111	2.830	2.456	8.009		
C12	2.778	3.505	7.717	12.285		
C13	2.778	2.032	7.717	4.129		
C14	2.556	2.110	6.533	4.452		
C15	3.111	2.110	9.678	4.452		
C23	2.244	2.418	5.036	5.848		
C24	3	2.275	9.000	5.176		
C25	2.333	2.908	5.443	8.456		

Optimization of the compressive strength of Palm Kernel Shell lightweight Aggregate concrete ..

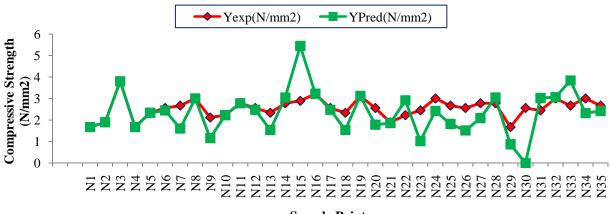
C34	1.733	1.855	3.003	3.441
C35	2.667	2.798	7.113	7.829
C45	2.711	2.010	7.350	4.040
Total	38.911	36.725	102.883	94.113

$$\begin{split} &\mathsf{N} = \text{total scores } = 30; \, \mathsf{K} = 2; \, \mathsf{Df}_b = \mathsf{K} - 1 = 1; \, \mathsf{Df}_w = \mathsf{N} - \mathsf{K} = 28\\ &\mathbf{SS}_b = \frac{(\Sigma(exp))^2}{15} + \frac{(\Sigma(pred))^2}{15} - \frac{((\Sigma(exp)) + (\Sigma(pred)))^2}{30}\\ = \frac{(38.911)^2}{15} + \frac{(36.725)^2}{15} - \frac{((38.911) + (36.725))^2}{30} = 190.853 - 190.693 = \textbf{0.16}\\ &\mathbf{SS}_w = \left(\sum(exp)^2\right) + \left(\sum(pred)^2\right) - \frac{(\Sigma(exp))^2}{15} + \frac{(\Sigma(pred))^2}{15} \\ = 102.883 + 94.113 - \frac{((38.911))^2}{15} + \frac{((36.725))^2}{15} = 196.996 - 190.853 = \textbf{6}.\textbf{143}\\ &\mathbf{MS}_b = \frac{\mathsf{SS}_b}{\mathsf{Df}_b} = \frac{0.16}{1} = \textbf{0.16}\\ &\mathbf{MS}_w = \frac{\mathsf{SS}_w}{\mathsf{Df}_w} = \frac{6.143}{28} = \textbf{0.2194}\\ &\mathsf{F} = \frac{\mathsf{MS}_b}{\mathsf{MS}_w} = \frac{0.16}{0.2194} = \textbf{0.7293}\\ &\mathsf{Fcrit} = 4.2582 \ (\mathsf{F}\ distribution table). \end{split}$$

Table 9: Summary of ANOVA					
Groups	Count	Sum	Average		
Expt.	15	38.911	2.5941		
Predict	15	36.725	2.4483		

ANOVA

Source of Variance	SS	df	MS	F	Fcrit	
Between Groups	0.16	1	0.16	0.7293	4.2582	
Within Groups	6.143	28	0.2194			
Total	6.303	29				



Sample Points

Figure 5.1 Experiment vs Model Compressive strength

VI. DISCUSSION OF RESULTS

Slump value averaged 0.00 mm for granite replacement rate of 100% using PKS. This low slump value may be due high water absorption capability hence taking up the required water content from the mix design, and more water will be required in such a situation in line with [29] The average density of concrete based on a 100% replacement was determined to be 1468 kg/m³ hence; the concrete can be classified as lightweight concrete [4, 35, 37, 42]. Using the equation produced by Scheffe's simplex model, the strength of light-weight aggregate concrete including palm kernel shells was optimized. The strengths of different mix ratios may be predicted by the model and vice versa. The batch with point (N15) and mix ratio of (0.5875: 1.0: 1.7375: 2.4625) for water, cement, fine aggregates, and PKS, respectively, has the highest predicted compressive

strength of 5.45 N/mm² according to the model results in tables 3.1 and 4.1. The laboratory values for compressive strength were found to be between (1.67 N/mm² - 3.8 N/mm²). Both the laboratory and model predicted compressive strengths are far lower than the minimum standard of 17.2 Mpa specified for structural lightweight concrete in [35, 42].

The low compressive strength in this research may be attributed to the lightweight, shapes and semiporous nature of PKS aggregate or breakdown of bonds between the aggregates and the paste, failure of shell aggregate and aggregate-paste interface as obtainable in [31]. The statistical tests used in this work demonstrated that Scheffe's model was sufficient for maximizing the compressive strengths of the palm kernel shell concrete. The value of F_{crit} , being greater than $F_{cal.}$ Indicates a good relationship between the experimental and modeled values. The derived model equations were found to be adequate for forecasting the strengths. The statistical tests conducted in this work demonstrated that Scheffe's model was sufficient for optimizing the compressive strengths of the palm kernel shell concrete. A WOLFRAM MATHEMAICAL computer program was used to select the optimized compressive strength of the palm kernel shells concrete as in Appendix I and vice - versa.

The statistical tests used here, proved the adequacy of the Scheffe's model for the optimization of the compressive strength of the palm kernel shells concrete and the compressive strength of all points in the simplex can be derived using this model. Because of the low compressive strength, the concrete produced in this research cannot be adopted for structural lightweight concrete construction.

VII. CONCLUSION:

The use of palm kernel shells in concrete production contributes in protecting our physical environment as it assists in preventing the depletion of the natural ground, means of waste disposal of the by-product to the areas of their production. Palm kernel shells are adequate for use in the production of light weight concrete. The responses (compressive strength) of the palm kernel shells aggregate concrete can be predicted by the mathematical model when the mix ratio is known or vice- versa.

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OPTIMIZATION MODEL

```
Maximize[
                                        \{eQnSt2, eQnSt2 = 3.8, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0\}, \{X_1, X_2, X_3, X_4\}\}
                               Minimize
                                        \{eQnSt2, eQnSt2 = 1.667, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0\}, \{X_1, X_2, X_3, X_4\}
 \label{eq:cond} \texttt{Conduct} = \{\texttt{3.80001}, \{\texttt{X}_1 \rightarrow \texttt{0.0859379}, \texttt{X}_2 \rightarrow \texttt{0.797634}, \texttt{X}_3 \rightarrow \texttt{0.10659}, \texttt{X}_4 \rightarrow \texttt{0.00983782}\}\}
 \texttt{Output} = \{\texttt{1.66699}, \{\texttt{X}_1 \rightarrow \texttt{0.0992361}, \texttt{X}_2 \rightarrow \texttt{0.782031}, \texttt{X}_3 \rightarrow \texttt{0.116365}, \texttt{X}_4 \rightarrow \texttt{0.00236768}\}\}
Cut = FittedModel || = 47 \times + \infty 1 \times + 38418.7 X_3 (= 1 \times)^2 X_4 + 9019.72 X_1 X_2 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_1 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_3 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_3 X_3 X_4^2 - 4748.66 X_2 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X_3 X_4^2 + 164110 X
                           Maximize
                                   \{eQnSt2, eQnSt2 = 17.4, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0\}, \{X_1, X_2, X_3, X_4\}\}
                           Minimize[
                                  \{eQnSt2, eQnSt2 = 1.67, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0\}, \{X_1, X_2, X_3, X_4\}\}
(17.4, \{X_1 \rightarrow 0.0785538, X_2 \rightarrow 0.803111, X_3 \rightarrow 0.101787, X_4 \rightarrow 0.0165481\}\}
(1.66999, \{X_1 \rightarrow 0.0992371, X_2 \rightarrow 0.782029, X_3 \rightarrow 0.116366, X_4 \rightarrow 0.00236785\})
 Cut(-)= FittedModel (47.2.4.3.4.1.2.4.3.4.1.2.4.3.4.1.2.4.3.4.4.4.9.019.72 X_1 X_2 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_1 X_3 X_2^2 - 4748.66 X_2 X_3 X_2^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3^2 - 4748.66 X_3 X_3 X_3 X_3^2 + 164 110 X_3 X_3 X_3 X_3 + 164 110 X_3 X_3 X_3 + 164 110 X_3 X_3 X_3 + 164 110 X_3 X_3 X_3 + 164 110 X_3 X_3 X_3 + 164 110 X_3 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3 + 164 110 X_3
                                                  Maximize[
                                                        \{eQnSt2, eQnSt2 = 23, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0\}, \{X_1, X_2, X_3, X_4 \ge 0\}
                                                  Minimize[
                                                       \{ eQnSt2, \ eQnSt2 = 20, \ X_1 + X_2 + X_3 + X_4 = 1, \ X_1 \ge 0, \ X_2 \ge 0, \ X_3 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3, \ X_3 = 1, \ X_1 \ge 0, \ X_2 \ge 0, \ X_3 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3, \ X_3 = 1, \ X_3 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_2, \ X_3 = 1, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_4 \ge 0, \ X_4 \ge 0, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_1, \ X_4 \ge 0, \ X_4 \ge 0, \ X_4 \ge 0, \ X_4 \ge 0, \ X_4 \ge 0 \}, \ \{X_4, \ X_4 \ge 0, \ X_
                     \texttt{Output} \quad \texttt{\{23., \{X_1 \rightarrow 0.0809488, X_2 \rightarrow 0.800047, X_3 \rightarrow 0.103504, X_4 \rightarrow 0.0155003\}\}}
                    (100) = \{20., \{X_1 \rightarrow 0.0805991, X_2 \rightarrow 0.800469, X_3 \rightarrow 0.103195, X_4 \rightarrow 0.015737\}\}
                        Maximize[
                                                        (eQnSt2, eQnSt2 = 37, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0), (X_1, X_2, X_3, X_4)]
                                                 Minimize(
                                                        (eQnSt2, eQnSt2 = 22, X_1 + X_2 + X_3 + X_4 = 1, X_1 \ge 0, X_2 \ge 0, X_3 \ge 0, X_4 \ge 0), (X_1, X_2, X_3, X_4)] 
                    (\texttt{37.0002}, \texttt{(X_3 \to 0.0325433, X_2 \to 0.71593, X_3 \to 0.023632, X_4 \to 0.227894)})
                    (22., \{X_1 \rightarrow 0.0805311, X_2 \rightarrow 0.800483, X_3 \rightarrow 0.103189, X_4 \rightarrow 0.0157972\})
```