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AN ESTIMATION OF WORKING CAPITAL MANAGEMENT ON PROFIT USING LOGISTIC REGRESSION AND DISCRIMINANT ANALYSIS

ABSTRACT

Aims: This paper estimates working capital management on profit using logistic regression and discriminant analysis on manufacturing and industrial firms in Ghana.

Study design: Research Paper.

Place and Duration of Study: Ghana, Secondary data for 2009 to 2014.

Methodology: Data in the form of ratios were computed from the audited annual financial reports of 13 manufacturing and industrial firms listed on the Ghana Stock Exchange covering the period from 2009 to 2014. This enables us to see whether a firm is profitable in a year or not.

Results: The results showed that the logistic regression of the dependent variable (Profit) on the independent variables such as Average Collection Period etc. was found to be significant and that there is no difference in variances for two firm classifications and this implies that a linear discriminant function $\hat{m}_{(2)}$ is effective in discriminating a firm who has managed its working capital on profit from one which did not.

Conclusion: This study showed that the binary logistic regression model estimates correctly at least 75% of firm's likelihood of managing working capital on profit while correctly discriminating the firms as having an effective management.

9
10 *Keywords: Binary logistic regression, working capital management, profitability, discriminant*
11 *analysis.*

1. INTRODUCTION

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15 Recently, management of working capital has been a major concern to financial economists
16 and accountants because it has an effect on the profit of firms. Economist and financial
17 analyst have over the years conducted research on the relationship between working capital
18 management and the profit of the firm. Chiou, Li and Han-Wen [4] defined working capital
19 management as the involvement of measures and policies by using firm's current assets and

20 liabilities in such a way that will sustain the working capital. Company's choice on policies to
21 manage firm's working capital have an effect on profitability.

22

23 Filbeck and Krueger [6] suggested that the progress of a firm depends on the ability of the
24 managers to effectively supervise inventories, receivables and payables. Management of
25 working capital in underdeveloped countries such as Ghana is imperative as creditors give
26 short-term credit to the long-term market. This may be due to the relatively higher inflation
27 rate in Ghana compared to other developed or developing countries [1]. Due to scarcity of
28 the finances of firms, there is the need for firms to effectively and efficiently manage their
29 working capital to gain profit.

30

31 Irfan [7] sampled 253 non- financial listed companies of Karachi Stock Exchange to
32 investigate the impact of working capital management on the performance of the firm in
33 Pakistan. The Ordinary Least Square Regression, Logistic regression and Pearson
34 Correlation techniques were used to analyze the results from the balance sheet of Stock
35 Listed Companies on Karachi Stock Exchange published by State Bank of Pakistan. He
36 identified from the result that only current asset over total sales from the five chosen
37 components of working capital management showed significant negative association
38 between working capital management and both proxies of performance which is return on
39 assets and return on equity whereas current asset over total asset, debtor's turnover,
40 inventory turnover and current ratio showed significant positive association with
41 performance. The results from the Logistic regression showed that current ratio, current
42 assets over total asset and current assets over total sales affect profit of firms.

43

44 Deloof [5] used correlation and regression analysis to examine a sample of 1,009 Belgian
45 non-financial firms for a period of 1992-1996 and identified a significant negative association
46 between gross operating income and the number of day's accounts receivable, accounts
47 payable and inventories of Belgian firms. He recommended that managers must minimize
48 the inventories and days of account receivable in order to increase profits of corporate firms.

49

50 Mathuva [10] used Pearson and Spearman's correlations, the pooled ordinary least square
51 and the fixed effects regression models to assess the effect of the components of working
52 capital management components on the profit of corporate firms using a sample of 30 firms
53 listed on the Nairobi Stock Exchange for the periods 1993 to 2008. He found that there a
54 highly significant negative association existed between accounts collection period and

55 profitability and a highly significant positive association between the inventory conversion
56 period, average payment period and profitability.

57

58 Almazari [3] analyzed a sample of 8 Saudi cement manufacturing companies listed in the
59 Saudi Stock Exchange for the period of 5 years from 2008-2012 to assess the association
60 between the working capital management and the firms' profitability. The study used
61 Pearson Bivariate correlation and regression analysis and discovered that current ratio had
62 the highest impact on profit and suggested that the cement firms must set a trade-off to
63 prevent liquidity or profit from being affected. He also identified that an increase in firm size
64 result in increase in profit. Also profit decreases when there is an increase in debt financing.
65 Hence a high correlation between the working capital management and profitability using
66 linear regression.

67

68 Akoto, Awunyo-Vitor and Angmor [1] investigated the correlation between working capital
69 management practices and profitability of listed manufacturing firms on Ghana Stock
70 Exchange using data collected from annual reports of all the 13 listed manufacturing firms in
71 Ghana covering the period from 2005-2009. Employing panel data methodology and
72 regression analysis, the study identified a significant negative correlation between
73 Profitability and Accounts Receivable Days. Also a significant positive correlation existed
74 between the firms' Current Asset Ratio, Current Asset Turnover, Cash Conversion Cycle,
75 Size and profitability of firms. The study recommended that incentives need to be generated
76 by managers to minimize accounts receivable to 30 days to create value for their
77 shareholders.

78 **2. METHODOLOGY**

79 We propose a methodology that seeks to address the problem of inadequate response on
80 working capital management which is clear in most stock exchange all over the world. The
81 second seeks to discriminate firms with regards to them having effective working capital
82 management on profit.

83 **2.1 Estimation of Working Capital Management on Profit**

84 Suppose D_i is the i th firm working capital management on profit for t years and K_i is
85 the i th firm's profitability in a year (nominal) for 6 years. Then on average, the i th firm
86 works $h_i = \frac{K_i \times t}{6}$ ($i = 1, 2, 3, \dots, n$) on profit in t years, where n is the total number of firms
87 who had acquire (non-zero) profit in the last t years at the time of study. Next, let

88 $R_i = \frac{D_i}{h_i}, i = 1, 2, \dots, n$ then a firm is said to have managed its working capital on profit if

89 $R_i > m$, where $0 < m < 1$

90

91 For the i th firm, we define a dichotomous variable y_i such that $z_i = \begin{cases} 1 & \text{if } R_i > m \\ 0 & \text{otherwise} \end{cases}$, where m

92 is the threshold. Menard [9] examined the use of the binary logistic regression for a
93 dichotomous outcome variable with covariates which are statistically significant to show the
94 likelihood of belonging to any of the two categories

95 Now suppose the binomial logistic regression model:

$$96 \quad \log \text{it}(\lambda_i) = \beta_0 + \beta_1 X_1 + \dots + \beta_k X_k + \varepsilon \quad (1)$$

97 $\lambda_i = P(y_i = 1 / X_1, X_2, \dots, X_k)$ is significant and correctly classifies at least 75% of firms
98 who have managed their working capital on performance (Alan, 2012), then an estimate of
99 the model can be written as:

$$100 \quad \log \text{it}(\hat{\lambda}_i) = \hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_k X_k \quad (2)$$

101 where $\hat{\beta}_i, i = 0, 1, \dots, k$ are estimates of the parameters β_i and $\hat{\lambda}_i$ is the estimate of the
102 likelihood of a firm managing its working capital on profit.

103 Given the explanatory variables $X_1, X_2, \dots, X_k, \lambda_i$ can be estimated as follows:

$$104 \quad \hat{\lambda}_i = \frac{\exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_k X_k\}}{1 + \exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_k X_k\}} \quad (3)$$

105 Model (3) is then used to estimate the likelihood of a firm's working capital management on
106 profitability for all firms observed to have worked to gain profit in the past t years.

107 These estimates are then used to generate a discriminant model (function).

108 **2.2 Discriminant models (functions)**

109 Suppose the random variable λ has probability density functions $f_1(\lambda)$ and $f_2(\lambda)$ for the
110 populations θ_1 (firms that managed their working capital on profit) and θ_2 (firms that did
111 not manage their working capital on profit), respectively. A firm whose likelihood π of
112 managing working capital on profit must be assigned to either θ_1 or θ_2 .

113

114 Let Ω denote the sample space of λ and A_1 and $A_2 = \Omega - A_1$ form a partition of Ω . If
 115 A_1 is the set of all values of λ for which a household is classified as θ_1 and A_2 is the set
 116 of values of λ for which a household is classified as θ_2 , then the (conditional) probability of
 117 classifying a population as belonging to θ_1 when actually it belong to θ_2 is given by;

118 $P(2|1) = P(\lambda_{ij} \in A_2 | \lambda_{ij} \in \theta_1) = \int_{A_2} f_1(\lambda) d\lambda$ and the (conditional) probability of classifying

119 a population as belonging to θ_2 when actually it belongs to θ_1 is;

120 $P(1|2) = P(\lambda_{ij} \in A_1 | \lambda_{ij} \in \theta_2) = \int_{A_1} f_2(\lambda) d\lambda$. According to Johnson and Wichern [8] a

121 reasonable classification rule should have an expected cost of misclassification (ECM) as
 122 small as possible with:

123
$$ECM = c(2|1)P(2|1)P(1) + c(1|2)P(1|2)P(2) \quad (4)$$

124 Where, $c(i|j)$ is the cost of misclassifying a
 125 population θ_j as θ_i , ($i = 1, 2$) and $P(i)$, ($i = 1, 2$) is the prior probability of θ_i and
 126 $P(1) + P(2) = 1$

127 The regions A_1 and A_2 that minimize the ECM, according to Johnson and Wichern [8] are
 128 defined by the values λ for which the following holds:

129
$$A_1 : \frac{f_1(\lambda)}{f_2(\lambda)} \geq \frac{c(1|2)P(2)}{c(2|1)P(1)}$$

 130
$$A_2 : \frac{f_1(\lambda)}{f_2(\lambda)} < \frac{c(1|2)P(2)}{c(2|1)P(1)} \quad (5)$$

131 According to Johnson and Wichern [8], If θ_i ($i = 1, 2$) has a normal distribution with mean μ_i
 132 and variance σ_i^2 , then the density ratio based on λ is given by:

133
$$\frac{f_1(\lambda)}{f_2(\lambda)} = \frac{\frac{1}{\sqrt{2\pi}\sigma_1} \times e^{-\frac{1}{2}\left(\frac{\lambda-\mu_1}{\sigma_1}\right)^2}}{\frac{1}{\sqrt{2\pi}\sigma_2} \times e^{-\frac{1}{2}\left(\frac{\lambda-\mu_2}{\sigma_2}\right)^2}}$$

$$134 \quad = \left[\frac{\sigma_2}{\sigma_1} \right]^{\frac{1}{2}} e^{-\frac{1}{2} \left[\left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_2^2} \right) \lambda^2 - 2 \left(\frac{\mu_1}{\sigma_1^2} - \frac{\mu_2}{\sigma_2^2} \right) \lambda + \left(\frac{\mu_1^2}{\sigma_1^2} - \frac{\mu_2^2}{\sigma_2^2} \right) \right]} \quad (6)$$

135 Rearranging and taking the natural logarithm of both sides, the first inequality in (3), by trivial
136 algebra becomes:

$$137 \quad \frac{1}{2} (\sigma_1^2 - \sigma_2^2) \lambda^2 + (\mu_1 \sigma_2^2 - \mu_2 \sigma_1^2) \lambda + (\mu_1^2 \sigma_2^2 - \mu_2^2 \sigma_1^2) \geq \sigma_1^2 \sigma_2^2 \ln \sigma_1 \sigma_2 c(1|2)P(2)c(2|1)P(1)$$

138 However, if $\sigma_1 = \sigma_2 = \sigma$ Eq. (6) becomes:

139

$$140 \quad \frac{f_1(\lambda)}{f_2(\lambda)} = \left[\frac{\sigma_2}{\sigma_1} \right]^{\frac{1}{2}} e^{\frac{(\mu_1 - \mu_2)\lambda}{\sigma^2} - \frac{1}{2\sigma^2}(\mu_1^2 - \mu_2^2)} \quad (8)$$

141 Again re-arranging and taking the natural logarithm of both sides, the first inequality of (5)
142 becomes:

143

$$144 \quad (\mu_1 - \mu_2)\lambda - \frac{1}{2}(\mu_1^2 - \mu_2^2) \geq \sigma^2 \ln \left[\frac{c(1|2)P(2)}{c(2|1)P(1)} \right] \quad (9)$$

145 Now, labeling the left hand side of (7) and (9) as quadratic and linear discriminant functions
146 $m_{(1)}$ and $m_{(2)}$ and the corresponding right hand sides as the critical values $c_{(1)}$ and $c_{(2)}$
147 respectively, the sample estimate of the discriminant functions and their critical values are
148 given by:

149

$$150 \quad \hat{m}_{(1)} = \frac{1}{2} (s_1^2 - s_2^2) \lambda^2 + (\bar{\lambda}_1 s_2^2 - \bar{\lambda}_2 s_1^2) \lambda + (\bar{\lambda}_1^2 s_2^2 - \bar{\lambda}_2^2 s_1^2) \quad (10)$$

$$151 \quad \text{with } \hat{c}_{(1)} = s_1^2 s_2^2 \ln \left[\left(\frac{s_1}{s_2} \right)^{\frac{1}{2}} \frac{c(1|2)P(2)}{c(2|1)P(1)} \right]$$

152 and

153

$$154 \quad \hat{m}_{(2)} = (\bar{\lambda}_1 - \bar{\lambda}_2)\lambda - \frac{1}{2}(\bar{\lambda}_1^2 - \bar{\lambda}_2^2) \quad (11)$$

$$155 \quad \text{with } \hat{c}_{(2)} = s^2 \ln \left[\frac{c(1|2)P(2)}{c(2|1)P(1)} \right]$$

156 where, $\bar{\lambda}_i = \frac{1}{n_i} \sum_{j=1}^{n_i} \lambda_{ij}$ and $s_i^2 = \frac{1}{n_i - 1} \sum_{j=1}^{n_i} (\lambda_{ij} - \bar{\lambda}_i)^2$ are based on samples of size n_i from

157 population $\theta_i (i = 1, 2)$; and $s^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$ is the pooled sample variance.

158 By the minimum ECM rule, a firm with λ likelihood of managing profit is classified as
 159 managing working capital on profit if:

160 $\hat{m}_{(2)}(\lambda_{ij}) = \frac{1}{2}(s_1^2 - s_2^2)\lambda_{ij}^2 + (\bar{\lambda}_1 s_2^2 - \bar{\lambda}_2 s_1^2) + (\bar{\lambda}_1^2 s_2^2 - \bar{\lambda}_2^2 s_1^2) \geq \hat{c}_{(2)}$, For $\sigma_1 \neq \sigma_2$ or

161 $\hat{m}_{(1)}(\lambda_{ij}) = (\bar{\lambda}_1 - \bar{\lambda}_2)\lambda_{ij} - \frac{1}{2}(\bar{\lambda}_1^2 - \bar{\lambda}_2^2)$, for $\sigma_1 = \sigma_2$ (12)

162 The discriminant functions $\hat{m}_{(1)}$ and $\hat{m}_{(2)}$ are effective in classifying a firm as having
 163 managed working capital on profit or not if μ_1 is significantly different from μ_2 .

164 **3. RESULTS AND DISCUSSION**

165

166 To apply the methodology proposed by this study, data in the form of ratios were computed
 167 from 13 manufacturing and industrial firms listed on the Ghana Stock Exchange covering the
 168 period from 2009 to 2014. Data which were the audited annual financial reports were
 169 collected from the Fact Book of the Ghana Stock Exchange and the web portals of the firms.
 170 This enables us to see whether a firm is profitable in a year or not. The dependent variable
 171 used in this study was Profitability. In order to analyze the effects of working capital
 172 components on the profitability of manufacturing companies in Ghana, profitability is
 173 measured by Return on Assets (ROA), which is defined as the ratio of earnings before
 174 interest and tax to total assets. Management of working capital was assessed for 6
 175 years (*i.e.*, $k = 6$).

176

177 For each firm, 6 years of non-zero profit z_i was computed from the nominal non-zero
 178 working capital. The proportion R_i of non-zero working capital accounted for by working
 179 capital on profit was computed for each of the 13 manufacturing and industrial firms who
 180 have managed their working capital on profit. The study used a threshold value of $m = 0.05$;
 181 and so all firms with $R_i > 0.05$ were classified to have managed their working capital on
 182 profit. Thus, the dependent variable for the binary logistic regression is defined by $Y_i = 1$ for
 183 managing working capital on profit and $Y_i = 0$, otherwise. After multicollinearity diagnosis,

184 Average Collection Period(X_1), Inventory Conversion Period(X_2), Average Payment
 185 Period(X_3), GROWTH(X_4), Debt Ratio(X_5), Current Ratio(X_6) and Company Size
 186 (X_7) were the explanatory variables used in setting up the logistic regression model.

187

188 The logistic regression of the dependent variable (Y_1) on the above variables was found to
 189 be significant ($\chi^2 = 376.206, df = 7, p - value < 0.001$). Table 1 shows the estimates of
 190 the fitted binary logistic regression model for the data. The fitted logistic regression equation
 191 is given by:

192
$$\log it(\hat{\pi}) = -2.035 - 0.021 X_1 + 0.002 X_2 + 0.008 X_3 + 0.147 X_4 - 0.62 X_5 + 0.032 X_6 + 0.09 X_7$$

193

194 This implies that:

195
$$\hat{\pi}_i = \frac{\exp\{\hat{\beta} X'\}}{1 + \exp\{\hat{\beta} X'\}}$$

196 Where,

197
$$\hat{\beta} = (-2.035 - 0.021 + 0.002 + 0.008 + 0.147 - 0.62 + 0.032 + 0.09)'$$
 and $X = (1, x_1, x_2, x_3, x_4, x_5)'$.

198

199 Equation (11), was used to compute the likelihood $\hat{\lambda}$ of managing working capital on profits
 200 for 13 manufacturing and industrial firms.

201

202

203 Table 1: Fitted binary logistic regression

Variable	β	SE	Odds ratio	p-value
Constant	-2.035	0.352	0.421	0.000
ACP	-0.021	0.019	0.754	0.000
ICP	0.002	0.027	0.692	0.000
APP	0.008	0.0035	1.254	0.000
Size	0.147	0.024	0.932	0.000
DR	-0.62	0.157	1.48	0.008
CR	0.032	0.013	0.833	0.000
Growth	0.09	0.017	1.458	0.003

204 Source: Authors' computation using GSE

205

206 Table 1 shows the statistically significant binary logistic regression model for the
 207 manufacturing and industrial data. The variables Average Collection Period, Inventory
 208 Conversion Period, Average Payment Period , Growth, Debt Ratio, Company Size and
 209 Current Ratio was found to be significant at determining the likelihood of managing working
 210 capital on profit for firms captured in the data.

211

212 Table 2: Frequency distribution for classification of 13 manufacturing and industrial firms for
 213 6 years making 78 observations with reported non zero- profit.

Classification	N	Mean	S.D	S.E
Effective Management	69	0.27	0.067	0.033
Non-Effective Management	9	0.34	0.022	0.0017
Total	78			

214 Source: Author's computation using GSE

215

216 Table 2 displays the descriptive statistics of likelihood of an effective management of a firm's
 217 working capital on profit. The Levene's test for the equality of variance of likelihood of
 218 managing a firm's working capital on profit by the two groups of firms was not significant ($F=$
 219 0.782 , p -value = 0.326). Hence, there is no difference in variances for two firm classifications
 220 and this implies that a linear discriminant function $\hat{m}_{(2)}$, Eq.(10) is appropriate.

221 Based on the data, an estimate of the common variance called the pooled variance for the
 222 two groups of households was found to be 0.004017 . And hence the linear discriminant
 223 function for the likelihood of managing a firm's working capital on profit is given by;

224
$$\hat{m}_{(2)} = (\hat{\lambda}_1 - \hat{\lambda}_2)\lambda - \frac{1}{2}(\hat{\lambda}_1^2 - \hat{\lambda}_2^2) = 0.042\lambda - 0.006219$$
 and

225
$$\hat{c}_{(2)} = s^2 \ln \left[\frac{c(1|2)P(2)}{c(2|1)P(1)} \right] = 0$$
 , on the assumption of equal cost of misclassification and

226 equal prior probabilities for both groups of firms. Therefore a firm with π likelihood of
 227 managing profit is said to have managed working capital on profit if $\hat{m}_{(2)} > 0$. The
 228 independent sample t-test for equal mean likelihood of managing a firm's working capital on
 229 profit is significant ($t = 12.625$, $df = 76$, p -value <0.001) and so the $\hat{m}_{(2)}$ is effective in
 230 discriminating a firm who has managed its working capital on profit from one which did not.

231 This result is applied to the whole 13 manufacturing and industrial firms irrespective of
232 whether or not a firm reported managing its working capital on profit after estimating the
233 likelihood of capital management for each of the firm in the entire data set.

234 **4. CONCLUSION**

235 This study proposes a methodology for analyzing the management of working capital on
236 profit in statistically underdeveloped countries. A binary logistic regression model, based on
237 data from firms with reported non-zero working capital on profit, is proposed for the
238 estimation of the likelihood of working capital on profit for all firms irrespective of whether
239 they managed their working capital on profit or not. "Univariate "discriminant functions, also
240 based on data from firms who managed their working capital on profit within the reference
241 period of Ghana Stock exchange, were proposed for discriminating firms that made effective
242 management of working capital on profit from those who did not.

243

244 An application of this methodology to the data from the Ghana Stock Exchange indicates
245 that the binary logistic regression model estimates correctly at least 75% of firm's likelihood
246 of managing working capital on profit while correctly discriminating the firms as having an
247 effective management.

248

249 A validation work should be conducted on the model using sample data from other sectors
250 on the Ghana stock exchange to further affirm the strength of the model and to show its
251 structural stability over time. However, this study clearly show that the methodology being
252 proposed is efficient in classifying all firms as having effectively managed its working capital
253 on profit or otherwise.

254

255 It is recommended that, the government should enforce that policies are implemented to
256 safeguard manufacturing and construction firms in Ghana and restrict importation of goods
257 into the country to boost the demand for goods or materials that are manufactured in Ghana
258 both in the short or long run.

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