Mathematical Modeling of Coronary Artery Disease (CAD): Analysis Reveals HbA1c and Total Cholesterol to be Significant Risk Predictors

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Abstract The increasing prevalence of CAD (Coronary Artery Disease) calls for early detection of risk factors and effective clinical management. The predictive potential of commonly estimated clinical variables on CAD incidence was assessed using mathematical modeling and analysis. A random sample of 50 patients with CAD and a control group of 50 subjects without CAD were drawn from a cardiac specialty hospital in Chennai, India during 2011-2012 (mean age = 50.2 years, SD = 11.2 years). Medical data included age, gender, height, weight, body mass index, presence/absence of hypertension, systolic blood pressure, diastolic blood pressure, presence/absence of diabetes mellitus, fasting blood sugar, post-prandial blood sugar, HbA1c, total cholesterol, family history of CAD. Mathematical modeling using discriminant analysis was performed to understand significant contributors leading to CAD. The discriminant analysis resulted in a mathematical model using parameters, HbA1c and cholesterol. The model was found to be statistically significant and this was demonstrated by computing the F value. HbA1c and total cholesterol were found to be significant in predicting the occurrence of CAD.

Keywords Mathematical Modeling, CAD, HbA1c, Cholesterol, Hypertension, Risk Factors, Discriminant Analysis

1. Introduction

In the emerging healthcare scenario in developing countries, urbanized lifestyle changes are resulting in increased prevalence of cardiovascular risk factors (diabetes, obesity, hypertension, and hypercholesterolemia). These in turn result in cardiovascular diseases (CVD) that include coronary heart disease and stroke [1].

Ferrari et al [2] found that the prevalence of some of the important CVD risk factors in CAD varied quite significantly across the seven geographical zones (Western / Central Europe; Canada / South Africa / Australia / UK; Eastern Europe; Central / South America; Middle East; East Asia; and India) categorized in a 5-year observational longitudinal cohort study – CLARIFY. Important observations on this varied prevalence were: (i) obesity ranged from 20% (East Asia) to 42% (Middle East), (ii) hypertension from 28% (Central/South America and East

It is well recognized that diabetes with other co-morbidities places the patient at potential risk of developing CVD. Still, there is greater need to step-up medical management of co-morbidities - hypertension and hyperlipidemia in patients with type 2 diabetes [3]. Avogaro et al., [4] reported that hyperglycemia and hypertension (particularly in men), and diabetic dyslipidemia (particularly in women), are risk factors of coronary heart disease (CHD) that can be modified with aggressive medical treatment. These risk characteristics point to vital implications in clinical management, wherein effective controls of glycemia, hypertension, and hyperlipidemia have been suggested to prevent CHD in diabetic patients. Diabetes induces a higher CHD risk in women, and higher ischemic stroke risk in men [5].

Interestingly, the link between glucose levels and

Asia) to 48% (Eastern Europe), (iii) increased LDL cholesterol levels from 24% (Canada/South Africa/Australia/UK) to 65% (Eastern Europe), (iv) increased heart rate (≥70 bpm) from 38% (Western/Central Europe) to 78% (India), (v) diabetes from 17% (Eastern Europe) to 60% (Middle East), and (vi) smoking from 6% (Central/South America) to 19% (Eastern Europe).

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cardiovascular disease/risk is not limited to diabetic patients. In fact, it is reported that non-diabetic individuals with higher baseline blood glucose levels (after adjustment for covariant) may carry a significantly high risk for developing CAD [6]. Kim et al [7] recently reported that when the mean HbA1c levels were above 6.2%, the CAD risk was higher and that HbA1c level can be one of the predictors for CAD even in the absence of diabetes.

In the present study, we have used discriminant analysis to model and determine the biochemical variables that show close correlations in the development of CAD. Clinical values of common risk parameters such as glycosylated hemoglobin, glucose, cholesterol, blood pressure, body mass index and body weight were used to develop the model.

2. Study Design and Methodology

Medical data of 50 patients with CAD and a control group of 50 people without CAD were collected from a cardiac specialty hospital in Chennai, India between October 2011 and June 2012. The average age of the cohort was 50.2 years with a standard deviation of 11.2 years. The study cohort characteristics were as follows: *Group I*: Patients with abnormal values of risk factors with Coronary Artery Disease (Male: Female ratio was 44: 24, average age was 53.4 ± 9.77 years (age range: 32-78 years); *Group II*: Subjects with controlled risk factors/ no risk factors without Coronary Artery Disease (Male: Female ratio was 39: 14, average age was 46.04 ± 11.54 years (age range: 21-75 years). The study was approved by the Institutional Ethics Committee of Frontier Lifeline Hospital.

CAD means patients with acute coronary syndrome (ACS) and / or positive Treadmill test (TMT) and / or angiographically proven CAD. Without CAD means patients with negative TMT or normal coronaries by angiogram (< 6 months ago). The inclusion criteria included all patients who were diagnosed with CAD during the last six months. Patients with symptoms of chest pain with or without risk factors but not proved either by ACS or TMT or Coronary Angiogram were excluded from the study. Further, patients who have been diagnosed with CAD over a period greater than six months were excluded from the study as this group would have been on medication for significant period of time which would have resulted in impacting the medical parameters.

The risk factors taken into consideration for the study were weight, body mass index, presence/absence of hypertension, systolic blood pressure, diastolic blood pressure, presence/absence of diabetes mellitus, fasting blood sugar, post prandial blood sugar, HbA1c, total cholesterol, and family history of CAD. Fasting total cholesterol > 200mg/dl was considered abnormal. A patient

was considered as uncontrolled diabetic if the fasting blood glucose > = 126 mg/dl or if the two hour (post-prandial) blood glucose level was > =200 mg/dl, or if the HbA1c (Glycosylated Hb) was 6.5% or above. Family history of premature CAD in first degree relatives was defined as IHD or sudden cardiac death before 55 years of age (in men) or 65 years of age (in women). BMI (Body mass index)> 30 was considered as obese. A patient was considered as hypertensive if the systolic blood pressure was 140 mm Hg and above or the diastolic blood pressure was 90 mmHg or higher [8].

Discriminant analysis was performed with all risk parameters. Each iteration resulted in the removal of the least significant variable. The statistical significance of the model was determined using the F values. The observed F value was compared with the tabulated F value and the statistical significance of the model was arrived at [9, 10].

3. Mathematical Modeling and Discussion

In a review of CHD and its risk burden in India, [11] pointed out that the measures should be implemented to screen for common risk factors such as high blood pressure, dyslipidemia, diabetes, and to provide for their mitigation in a cost-effective manner, in order to lessen the CHD deaths. Suitable lifestyle changes (including physical activity) and nutritious diet (balanced intake of fruits and vegetables) would complement these preventive steps in the control of CHD burden. Our previous mathematical model and analysis of CVD burden in diabetic population, clearly indicated the effective control of associated co-morbidities (hypertension and hyperlipidemia) substantially reduced the incidence of CVD [12].

In the present study, the routinely assessed clinical/biochemical parameters of each patient/ subject were analysed and correlated with the presence/absence of CAD. The modeling was carried out with a set of 50 patients and a set of 50 subjects without CAD as control. Step wise discriminant analysis led to equations (1) and (2).

$$10.65 \text{ K}_1 - 3171.7 \text{ K}_2 = -1.92$$
 (1)

$$-3171.7 K_1 - 20664.9 K_2 = -40.32$$
 (2)

Where K_1 and K_2 are the weights associated to the parameters HbA1c and total cholesterol. Solving the above equations, we get discriminant criteria Z where

$$Z = K1 * HbA1c + K2 * total cholesterol where K1 = 0.00063 and K2 = 0.000858.$$

The discriminant criteria value Z was computed for each person from the control group and for each patient with CAD which is presented in figure 1.

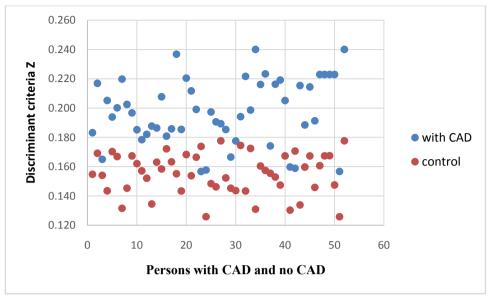


Figure 1. Discriminant criteria using parameters HBA1C and total cholesterol

From figure 1, we see that there is significant discrimination between the two groups, control and CAD group. This implies that the derived weights K_1 and K_2 are able to significantly discriminate the two groups. Statistical treatment of the analysis was carried out and the calculated F value was found to be 466.2. The tabulated $F_{2,97}$ (5% level) value was found to be 3.09 which is much lower than the calculated F value thereby indicating that the model derived is statistically significant. We also found that a score of 0.18 and above indicates presence of CAD while a score below 0.18 indicates absence of CAD. To validate this, we chose a test group of four patients with CAD and six persons without CAD (Table 1). The Z score cut-off clearly indicates that the model was able to delineate the presence or absence of CAD in the sample cases.

Table 1. Validation of CAD predictive potential of derived Discriminant score Z.

S. No.		HbA1c	Total cholesterol	Discriminant score Z
1	Patients with CAD	5.9	215	0.186
2		5.3	216	0.181
3		4.9	273	0.214
4		6.0	206	0.181
5	Subjects without CAD	5.1	195	0.166
6		6.0	183	0.167
7		6.4	146	0.147
8		6.1	123	0.109
9		4.9	160	0.140
10		6.2	180	0.158

From table 1, we find that the discriminant scores of patients with CAD are all above 0.18 and the scores of the persons without CAD are in the range of 0.11 to 0.17 (all below 0.18). Therefore, the model has predicted accurately, the presence/absence of CAD.

The glycosylated hemoglobin HbA1c (unlike blood

glucose levels that measures the levels of glucose at an instant of time) provides a long-term trend of the circulating glucose. This measure is clinically relevant to understand the risk of development of other complications related to diabetes. In the study by Kim et al. [7], fasting glucose level and HbA1c were assessed for their effect on atherosclerotic development. It is well known that other co-morbidities (lipid levels, blood pressure) also contribute to disease progression. In this context, the findings of the present mathematical analysis, strongly suggests that HbA1c along with cholesterol levels may be an excellent predictor of CAD. While other biochemical/clinical variables were also tested for association with CAD predictability, the combination of HbA1c and total cholesterol showed strongest correlation, and there was an unambiguous delineation between CAD and non-CAD group (Fig. 1).

In another study by Verdoia et al [13] that included 1703 non-diabetic patients, HbA1c was found to be significantly associated with hypercholesterolemia amongst other risk factors. They also found that HbA1c (and not fasting blood glucose levels) to be significantly linked with atherosclerosis (carotid intima-media thickness (C-IMT) and carotid plagues prevalence). Ashraf et al [14] reported that the increase in HbA1c levels coincided with significant increase in CAD prevalence, and their multivariate analysis revealed HbA1C as an independent CAD predictor. Their study implicated the potential of HbA1c for CAD risk stratification and assessment of its severity (independent of conventional cardiovascular risk factors, insulin resistance inflammatory markers).

Similar to the geographic variation in the prevalence of CAD risk factors mentioned earlier, Ferrari et al. [2] also found that the extent of treatment-control of the risk factors differed considerably across the world. Their key findings were: blood pressure control in hypertension varied from 47% (Eastern Europe) to 66% (Central/South America), glucose control in diabetes from 23% (India) to 51%

(Western/Central Europe and East Asia), LDL-cholesterol and dyslipidemia control from 32% (Eastern Europe) to 75% (Canada/South Africa/Australia/UK), heart rate <70 bpm from 22% (India) to 62% (Western/Central Europe), and heart rate ≤60 bpm in angina patients from 2% (India) to 29% (Canada/South Africa/Australia/UK and Central/South America) [15].

To sum up, with the current scenario of immense prevalence of CAD and associated risk factors (diabetes mellitus, hyperlipidemia), the public health policy should certainly include both early detection of individuals with risk factors and their treatment for risk-reduction, as well as early detection of clinical cases and implement secondary prevent prevention measures to their long-term complications [11]. In this context, the discriminant score approach based on the cohort studied here, may be an useful objective index that can be used for diagnosis as well as prognosis in the clinical management of CAD. The cohort studied here was a select group that visited this hospital, and is not representative of a cross-section of a population, nor does it include patients with sub-clinical state of CAD (with no symptoms). It is suggested that this approach may be extended to larger cohorts.

4. Conclusions

The results of our study indicate mathematical modeling using discriminant analysis serves to predict presence/absence of CAD. Here, CAD was found to be a function of total cholesterol and HbA1c. The model was successfully validated using a sample set of cases. This model can be used to predict presence of CAD and can be a useful tool for physicians.

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