

Futures markets development as a price risk strategy in Iran's dates

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ABSTRACT

Price risk in agricultural products has created main financial problems for agricultural producers. There are different ways and instrument to deal with these price risks or price volatility. This paper focuses on futures markets and calculates Hedge Ratio for dates. A Bivariate BEKK GARCH model, is used to determine time varying Hedge Ratios. The results show that BEKK BGARCH hedge ratio for dates is 0.7 which hedging performance is higher than the traditional one. The BEKK BGARCH hedge ratios provide 80% variance reduction, which is inferior to the constant OLS procedures. The results of this paper suggest that futures markets are good instruments for managing Dates producer price risk in Iran.

Key words: Hedge ratio; BGARCH; Dates.

JEL Classification: C₁, C₁₅, C₃₂, D₄.

INTRODUCTION

One of the most important questions in agricultural economic is why producers are not engaged in futures markets to manage price risk as predicted by theoretical and empirical literature [1]. Important question in price risk management issues is whether market instruments such as future contracts and future markets, can be effectively used to manage price risk. Transfer of risk is one of the main functions of the futures markets. Risks are transferred to those willing to bear them, as hedgers reduce their risk by paying a premium to speculators. For agricultural commodities, risk may occur due to drought, near record production, an increase in demand, a decrease in international production, etc. Producers usually face different layers of risks, which can be categorized in to three types in general. First, normal variation in production, which do not require any policy response and should be directly managed by producers as part of their normal business strategy. Second, catastrophic events, which are infrequent and beyond the capacity of producers or markets to manage, require government's involvement. To deal with these risks, government usually launches support programs. Third, market risks. Occur due to the changes in supply and demand, which are beyond the control of producers skills but they can manage the risks by using market tools. This is where insurance and future market comes in [1]. The price risks are caused by the change in supply and demand which is beyond producer's control [9]. With a tendency toward a decline in government support programs like direct subsidies programs, producers are forced to identify these risks and use private risk management strategies such as future contracts to alleviate losses by themselves [1]. Hedging by the agricultural producers generally involves selling the commodity futures because producers of the commodity want to lock in a price floor. Simultaneously speculators and investors looking to lock in a price ceiling are buying the contract. The commodity futures markets thus provide a means to transfer risk between persons holding the physical commodity (hedgers) and investors speculating in the market [6].

Dates market status and its price volatility

Dates represents a major role in Iran agricultural economic. Figure 1 shows that the production of dates was relatively stable over the last 20 years. There was a slight forward line in 2007. In 2010 Iran's dates yield was 63326.4 Hg/Ha. Dates product provides nearby 1.3% of total agricultural export value. In the years 2008, 2009, 2010, dates cash receipts contributed 101783, 55819, 134001 thousand\$ from its export. The average revenue from date's production contributed more than 7206 Billion Rial during 2010. Iran was the second largest date exporters with 2.8% of the worlds date supply in 2010.

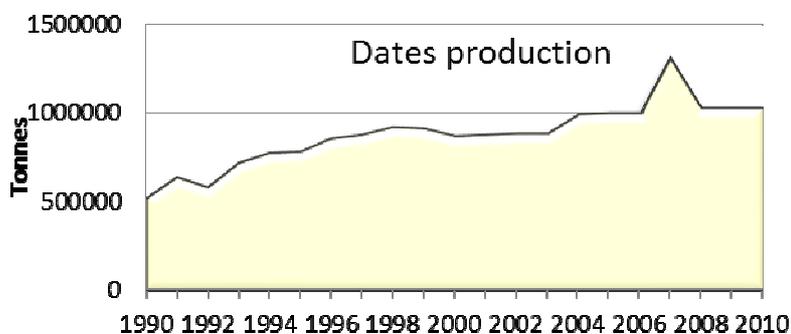


Figure 1: Dates production during 1990-2010

Trade is an important part of Iran's date's demand. According to F.A.O the amount of dates export was about 106760 Tones in 2010. Iran ranked as the second and forth largest dates exporters in the world for the past years. Iran has been the third largest Dates producers in the world in 2010. Iran's dates export has been fluctuated during 1990-2010. The maximum amount of dates export has been occurred in 1996. Figure 2 shows dates producers price in cash markets during 1990-2011. As figure 3 shows price was not stable over this time period. The difference between the highest price and the lowest price was almost 3354.3 Rials/KG during the past 20 years which shows that the highest dates producer's price was almost 2.5 times as much as the lowest price there fore date's price has had risk. Dates producers revenue is generated from the sale of dates. The correlation between dates producer price (PP) and date production with one lag (Y_{-1}) which is about -0.2 and the correlation between date market margin (MM) and date production with one lag (Y_{-1}) which is about -0.5 shows that there is fluctuation in the producer's and wholesaler's income (respectively) so dates price has risk.

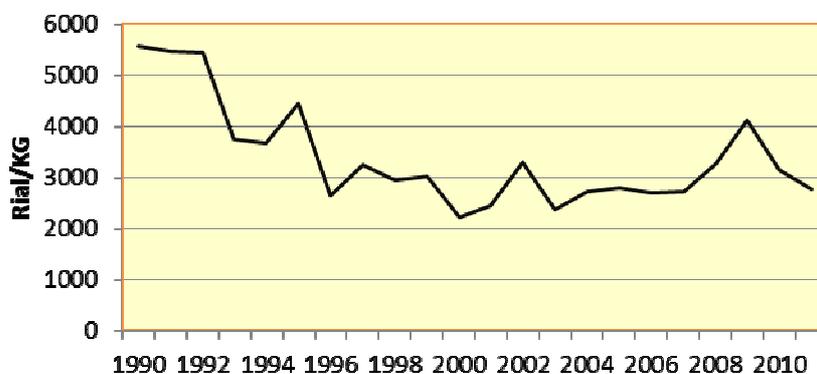


Figure 2: Iran's dates producer price during 1990-2011 in constant price 2004=100

A considerable amount of research has applied the Generalized Autoregressive Conditional Heteroscedasticity (GARCH) models proposed by Engle [8], Bollerslev [4,5] to estimate time-varying optimal hedge ratios. Di Ai [1] demonstrates optimal hedge ratios using futures contracts for Ontario and Alberta feedlot for live cattle and feeder cattle. This study estimated constant optimal hedge ratios using OLS and SUR models. Times varying optimal hedge ratios were estimated by GARCH models. Results indicated that time varying hedge ratios eliminated more risk than constant hedge ratios. Waldemar *et al* [11] determined optimal hedge ratios for soybean farmers in Rondonópolis, using the bivariate GARCH BEKK model, which considers the conditional correlations of the prices series, and then compared the results with the minimum variance model effectiveness. The financial effectiveness of the dynamic hedge model is superior and can be used by farmers for several decision making purposes. Taufiq Choudhry [6] investigated the hedging effectiveness of time-varying hedge ratios in the agricultural commodities futures markets using four different versions of the GARCH models. The GARCH models applied are the standard bivariate GARCH, the bivariate BEKK GARCH, the bivariate GARCH-X and the bivariate BEKK GARCH-X. Futures data

for corn, coffee, wheat, sugar, soybeans, live cattle and hogs are applied. Comparison of the hedging effectiveness is done for the within sample period (1980–2004), and two out-of-sample periods (2002–2004 and 2003–2004). Results indicated superior performance of the portfolios based on the GARCH-X model estimated hedge ratio during all periods. Kumar *et al* [10] examines hedging effectiveness of futures contract on a financial asset and commodities in Indian markets. They estimate dynamic and constant hedge ratio for S&P CNX Nifty index futures, Gold futures and Soybean futures. Various models (OLS, VAR, and VECM) are used to estimate constant hedge ratio. To estimate dynamic hedge ratios, they use VAR-MGARCH. It is found that in most of the cases, VAR-MGARCH model estimates of time varying hedge ratio provide highest variance reduction as compared to hedges based on constant hedge ratios. Baillie and Myers [2] examined six different commodities using daily data over two futures contract periods. They estimated Bivariate GARCH models of cash and future prices for six commodities. The models appear to fit the data well and estimated OHRs vary over time so that the usual assumption of constant OHRs is inappropriate.

MATERIALS AND METHODS

Data on Dates comprise monthly observations over the years 1990-2011 namely:

- Producer price in constant price 2004=100 in Rials of I.R Iran
- Wholesale price in constant price 2004=100 in Rials of I.R Iran
- Total Date product in Tones

All of above data are collected from the statistical office of central Bank of the I.R of Iran.

Time-Varying Hedge Ratios and Their Estimation Using BGARCH Models

- The Time-Varying Hedging Rule

Using a mean-variance framework, hedge ratios have been estimated using OLS by regressing the returns from holding a spot contract on returns from holding a futures contract. In a similar context, assuming utility maximization and efficiency in future markets, the conditional optimal one-period ahead hedge ratio, b_{t-1}^* , at time t can be derived as:

$$b_{t-1}^* = \frac{Cov(R_t^s, R_t^f | \Psi_{t-1})}{Var(R_t^f | \Psi_{t-1})} \quad (1)$$

Where R_t^s and R_t^f denote logarithmic of spot and futures prices from $t-1$ to t , respectively and Ψ_{t-1} is the information set at time $t-1$. This ratio is similar to the conventional hedge ratio except that the conditional variance and covariance replace their unconditional counterparts. Because conditional moments can change as the information set is updated, the hedge ratios also can change through time. [3].

The motivation behind using BGARCH model in the context of hedge ratio estimation is that, monthly date's future and spot prices react to the same information and have none zero covariance conditional of the available information set.

A natural and widely used model for estimating model 1 is BGARCH model. An alternative to the diagonal vech parameterization and appropriate bivariate model was tentatively specified as: [2]

$$\begin{aligned} \Delta y_t &= \mu + \varepsilon_t \\ \varepsilon_t | \Omega_{t-1} &\square N(0, H_t) \\ vech(H_t) &= C + A vech(\varepsilon_{t-1} \varepsilon'_{t-1}) + B vech(H_{t-1}) \end{aligned} \quad (2)$$

Where $y_t = (pp_t, ppf_t)'$ is a (2×1) vector containing cash and futures prices, H_t is a (2×2) conditional covariance matrix, C is a (3×1) parameter vector, A and B are (3×3) parameter matrixes, and $vech$ is the column stacking operator that stacks the lower triangular portion of a symmetric matrix. The diagonal vech parameterization involves nine conditional variance parameters.

RESULTS

Monthly cash (spot) and futures prices of dates are used in the empirical tests. Since dates are storable commodity so future price for a storable asset is considered equal to the assets cash price plus the assets cost of carry [7]. Because there are no future markets for dates in Iran, so in this paper, dates future prices is calculated with cost of carry method and also with forecasting methods (GARCH models and Artificial Neutral network techniques). Then, these three methods are compared with each other. All the data rage are from 1990:1-2011:12.

At first, Unit root tests for monthly date's producer prices in constant price 2004=100, is performed. Results of ADF test indicate that the null hypothesis could be rejected for spot price series at 5% significance level. So the producer price series (PP) and the logarithm of producer price series are stationary.

In the next step, ARCH-LM test is performed on date's spot price. According to AIC and SBC criteria two lags is accepted for spot price series, which equation is as follow: (standard deviation in parenthesis)

$$PP = 410 + 0.97PP_{-1} - 0.03PP_{-2}$$

(308) (0.14) (0.12) (3)

$$R^2 = 0.89 \quad DW = 1.99$$

PP is dates cash producer price. If there are ARCH effects left in residuals, the GARCH approach proved to be appropriate. Table number 1 shows that the null hypothesis could be rejected and ARCH effect is left. There fore GARCH model is appropriate to be used.

Table1: Results of ARCH-LM test for date's spot producer price model

ARCH test		probability
F-statistic	32.4	0.0
obs R-squared	29.03	0.0

Ref: research findings

Results of best GARCH (1, 1) is as follow: (standard deviation in parenthesis)

$$PP = 0.99PP_{-1} + 117100.7 + 0.6resid_{-1}^2 + 0.26GARCH_{-1}$$

0.003 48444 0.17 0.11 (4)

$$R^2 = 0.88 \quad DW = 1.96$$

After applying GARCH model, future prices series is forecasted using this GARCH model (from equation number 4). More over, future prices series is forecasted with artificial neural network (ANN) approach.

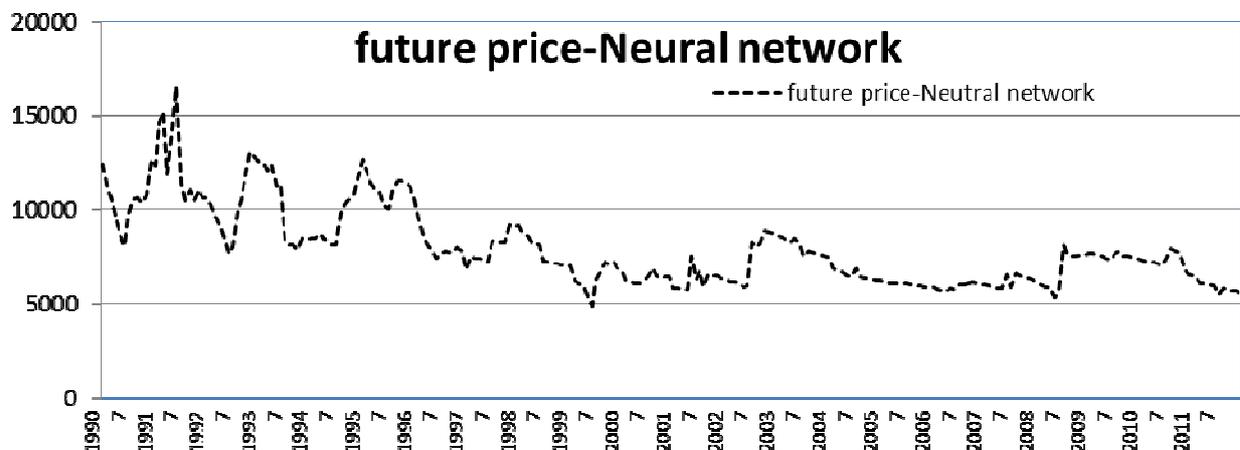


Figure 3: Monthly Dates future prices forecasted with different methods

This paper is used Multi-layer Perceptron Neural Network. Five neurons for the hidden layer and one neuron for output layer were chosen. In the hidden layer, logistic function as activation function and in the output layer, linear function as activation function was used. Figure number 3 shows the results of future prices forecasted for dates. After forecasting dates future prices with these three methods, according to minimum variance criteria, the ANN approach is preferred and outcome.

Determination of hedge ratios

The OLS regression was performed for date's prices. When the regression, $\log PP = c + \beta \log PPF$, was run, a trend variable was added for a better results. The coefficient of β is constant hedge ratio. Figure number 6 represents the OLS estimated constant hedge ratio from the minimum variance model. It is clear from equation number 16 that β coefficient is positive and significantly different from zero at 1% level. R^2 represents to what extent the variation of spot returns can be explained by futures returns [10]. Combining the optimal hedge ratios and R^2 values, the question that how much of the dates product should be hedged to maximize the risk reduction, can be answered. A date's producer who maintained 96% of dates hedged could reduce the date's price risk by 90%.

$$\begin{aligned} \log (PP) &= 0.33 + 0.96 \log (PPF) \\ (0.25) \quad (0.03) & \\ R^2 &= 0.9 \quad DW = 1.9 \end{aligned} \tag{5}$$

Table 2: Estimation of BGARCH models with BEKK

$$\begin{aligned} Model : \log S_t &= \hat{\mu}_1 + \varepsilon_{s,t} \\ \log F_t &= \hat{\mu}_2 + \varepsilon_{f,t} \end{aligned}$$

coefficient		
$\hat{\mu}_1$	8.7 (0.011)	Cash equation
$\hat{\mu}_2$	8.7 (0.01)	M ₃ =-0.07 M ₄ =1.9 J.B=15 Q(12)=0.5
\hat{M}_{11}	0.003 (0.001)	
\hat{M}_{22}	0.002 (0.001)	
\hat{M}_{12}	0.003 (0.0008)	
\hat{A}_{11}	0.6 (0.1)	Future equation
\hat{A}_{22}	0.8 (0.11)	M ₃ =-0.8 M ₄ =7.5 J.B=19 Q(12)=0.5
\hat{B}_{11}	0.5 (0.05)	
\hat{B}_{22}	0.39 (0.05)	
Log Likelihood	420	

Ref: research findings

Note: numbers in parenthesis are standard errors, m_3 is sample Skewness, m_4 is sample Kurtosis, J.B is Jarque-Bera normality test, Q(12) is the Box Pierce statistic for 12th order serial correlation in the residuals.

Table number 2 provides the parameter value for the BGARCH- BEKK results to examine the autoregressive conditional heteroskedasticity and interaction effects that exists for the PP and PPF. The ARCH coefficients are positive and significant; imply volatility in the log difference of cash price (A_{11}) and future price (A_{22}). Dates GARCH coefficient (B_{11} , B_{22}) are significantly positive implying GARCH effect. The covariance parameters indicate a positive and significant interaction between these two prices. The covariance GARCH parameters \hat{B}_{11} and \hat{B}_{22} , which account for the conditional covariance between cash and futures prices, are positive and

significant, imply strong interaction between cash and futures prices. The estimated Bivariate GARCH models with BEKK parameterization appear to provide a good represent of the data (LL=420). Since the main objective of this study was to determine the time varying optimal hedge ratios so this study used three above models (BEKK-Diagonal VECH- CCC) to determine BGARCH hedge ratios. According to minimum standard deviation hedge ratios and also to Hedge effectiveness criteria the BEKK is outperform. Figure number 4 shows time varying optimal hedge ratios for dates from BEKK model. The average of the BGARCH optimal hedge ratios was 0.7 which was slightly lower than the OLS constant optimal hedge ratio (0.96) during 1990-2011. Most of the movement of the time varying hedge ratios is confined below the constant minimum variance hedge ratio.

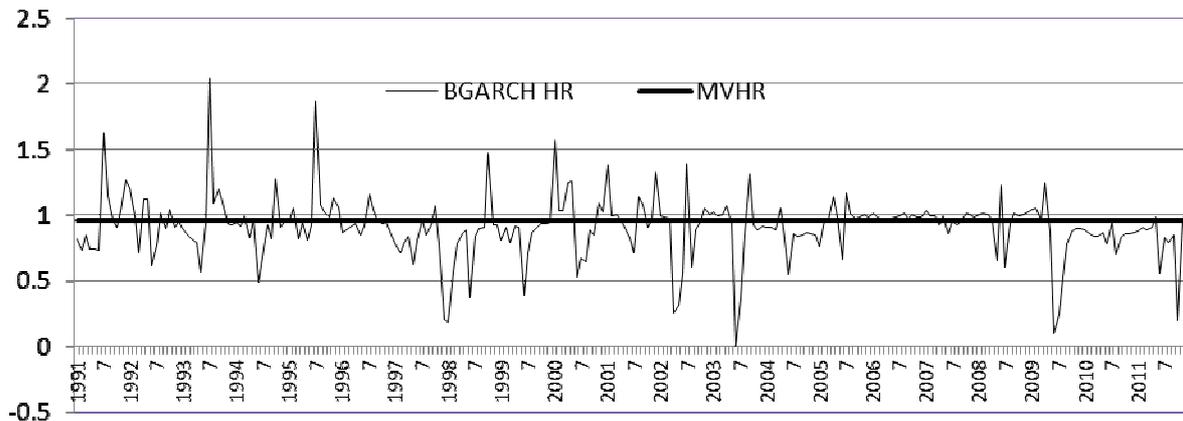


Figure 4: optimal hedge ratios from BGARCH model and OLS
(Time varying hedge ratios V.s constant hedge ratios)

CONCLUSION

In Iran the date's production represents a major role in agricultural economic. The date's producer price has been fluctuating over the last few decades. So these fluctuations effect on farmer's motivation and these fluctuations are forced farmers not to produce. In addition, Iranian government has gradually decreased the (in) direct subsidy paid to producers and forced the producers to take the primary responsibility for managing market risks. So this study examined for managing dates price risk by using futures markets. This study focuses on dates and studied hedging prices for dates. Since, there is not a future market for dates yet, so, we forecasted dates futures price with three different methods. Then, under the mean variance frame work, optimal hedge ratios were estimated. BGARCH models were applied to estimate time varying hedge ratios for dates during 1990-2011. This method permits the hedge ratios to be based on conditional information. Results of dates effectiveness show that the BGARCH hedge ratios perform better at reducing the variance of the portfolio return for dates in comparison to OLS hedge ratio. Hedging dates by using the BGARCH hedge ratio in future market was recommended based on the results of this study. Since there was not a big difference in risk reduction between using constant and time varying hedge ratios, both hedge ratios could achieve more than 70% price risk reduction, and dynamic hedge ratios is appeal.

One of the main functions of the futures market is to provide a hedging (risk transfer) mechanism. It is also a well-documented claim in the futures market literature that the optimal hedge ratio should be time-varying and not constant. So results of this paper show that the BGARCH hedge ratios outperform OLS. In general hedging dates in future markets is a good way and could eliminate producer's price risk to 80%.

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