THE CAUSALITY BETWEEN INFLATION AND UNEMPLOYMENT: 
THE INDONESIAN EVIDENCE

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Abstract. Two closely watched indicators of economic performance are inflation and unemployment. This study empirically analyzes the causality between inflation and unemployment in Indonesia during 1984 to 2017. The data were gathered from the Indonesian Central Bureau of Statistics. Methodologically, this study employed the Granger Causality test and Vector Autoregression to determine the causality between inflation and unemployment. The results show that there is a one-way causality between inflation and unemployment. The findings imply that unemployment causes inflation, but not vice versa. Next inflation and unemployment are also closely related to other determining factors, such as season, household income, and the decisions to attend school or to perform the housekeeping.

Keywords: granger causality, inflation, unemployment, vector autoregression.

JEL Classification: E600, E610.

Introduction

Post global financial crises (2008) have forced countries to adopt expansionary and stimulating macroeconomic policies aiming to reduce unemployment. Some countries, such as United Kingdom, Germany, and the United States of America have become successful in lowering the unemployment in their labor markets. However, Spain and Italy are stuck at high rates of unemployment with rigid labor markets (Bhattarai 2016). The unemployment could be stabilized towards their natural rates by stimulating the aggregate demand through fiscal or monetary policies with or without some increase in price levels (Keynes 1936, Phillips 1958, Benati 2015, Blanchard 2016).

Short-term economic problems, such as inflation and unemployment are among the most notable macroeconomic problems all the time (Al-zeaud 2014, Arshad 2014, Bhattarai 2016, Caporale and Škare 2011, Cioran 2014, Furuoka 2007, Furuoka 2008, Israel 2015, Katria et al., 2011, Kogid et al. 2011, Mahmood et al. 2013, Okafor et al. 2016, Sa’idu and Muhammad 2015, Ştefan and Bratu 2016, Thayaparan 2014, Touny 2013, Umaru and Zubairu 2012, Zaman et al. 2011, Pallis 2006, Benati 2015, Blanchard 2016). The Indonesian government started to focus on inflation when Indonesia experienced an economic shock during the transition period (1965–1969). Fortunately, the Indonesian government managed to control the inflation rate as Indonesia only had an inflation rate of below 10% in 1969 (Bank Indonesia 2004). However, the monetary crisis hit Indonesia again in 1997–1998 that resulted in the inflation rate of 58.4%. During the post-monetary crisis period, Indonesia managed to recover that caused the inflation rate to be below two digits. Further, the global financial crisis hit the global economy in 2008, but the Indonesian inflation rate remained stable. One of the likely factors of this condition is the government’ various economic rescue programs such as the tight money or contractive policy that was effective in taming the inflation rate. Besides, the Inflation Targeting Framework that was implemented by Bank Indonesia (the Indonesian central bank) since July

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findings suggest that the Phillips curve does not apply in the Philippines. The socioeconomic factors such as the output gaps likely explain the Phillips curve better in the Philippine context. In Nigeria, Umaru and Zubairu (2012) indicate that there is no causality between inflation and unemployment. The findings suggest that the Phillip curve does not apply in Nigeria and it is necessary to use the unemployment or inflation theory that is based more on the Nigerian data and situation.

Besides studies that show no causal relationship between inflation and unemployment, Caporale and Skare (2011) demonstrate that there is a one-way causal relationship between inflation and job opportunities. Their findings, based on the study on the Organisation for Economic Co-operation and Development (OECD) countries, thus suggest that inflation affects job opportunities, but not the way around. This condition recommends policymakers to pay more attention to the short-term and long-term job and output growth. In Malaysia, Furuoka (2007) also finds the one-way causality between inflation and unemployment, implying that inflation leads to unemployment but not the way around. The study also demonstrates the cointegration and causal relationship between inflation and unemployment in Malaysia. In other words, the results confirm the existence of the Phillips curve. Still in the same country, Kogid et al. (2011) document the one-way causality between inflation and unemployment. Their findings imply that inflation causes unemployment, but not vice versa. The results also confirm the trade-off relationship between inflation and unemployment in Malaysia and the government needs to ensure that the economic policies will facilitate sustainable economic growth in the future. Using the US data, Ştefan and Bratu (2016) also find the one-way relationship between inflation and unemployment, suggesting that inflation explains unemployment but not vice versa. Their findings suggest that policymakers should develop programs that reduce unemployment such as productive labor projects while at the same time also control inflation. Besides, the programs should focus on replacing foreign labors with local labors and on ensuring that the aggregate demands reach the optimal unemployment and inflation levels that will eventually support the long-term economic growth. The Pakistani study of Mahmood et al. (2013) demonstrate the one-way causality between inflation and unemployment, implying that inflation affects unemployment but not the way around. The study also suggests that the Pakistani policymakers maintain the equilibrium point between inflation, unemployment, and interest rate to control for economic shocks. Lack of focus on one of the three variables likely affects the economy. Still using the Pakistani data, Zaman et al. (2011) find the long-term relationship between inflation, unemployment, and interest rate to control for economic shocks. Lack of focus on one of the three variables likely affects the economy. Still using the Pakistani data, Zaman et al. (2011) find the long-term relationship between inflation and unemployment, denoting that inflation causes unemployment but not vice versa. The results also indicate that increasing inflation likely increases employment opportunities that eventually facilitates growth. The study empirically confirms the existence of the Phillips curve in Pakistan, both in the short-term and in the long-term. Nigeria also exhibits the one-way causality between inflation and unemployment by Sa’idu
and Muhammad (2015) that indicating inflation leads to unemployment but not the other way around. Their results recommend the joint efforts of all policymakers to restructure the economy to manage price instability and improve the infrastructures.

Further, Katria et al. (2011) who analyze the South Asian Association for Regional Cooperation (SAARC) countries, find a negative relationship between inflation and unemployment. Their results indicate that the collaboration between monetary and fiscal policies manages to stabilize the business cycle. Next, the Nigerian study by Okafor et al. (2016) indicates that inflation negatively affects unemployment. Their results recommend that policymakers not only rely on monetary targets but also on output targets through the economy deepening to maintain the optimal inflation rate and the minimal unemployment level. In a similar vein, Cioran (2014) demonstrates that inflation negatively affects unemployment in Romania and the European Union (EU). The findings suggest that inflation rate is an effective instrument to prevent increasing unemployment in the EU and Romania.

Besides the negative results, other studies find the positive relationship. For example, using the Egyptian data, Touny (2013) documents that unemployment positively affects inflation in the long run. The results recommend that policymakers implement their monetary policies to overcome the inflationary pressure regardless of the negative effects of unemployment. Further, Israel (2015) who analyzes several developed countries such as France, Germany, the UK, and the US, show the long-term positive relationship between inflation and unemployment. This positive relationship is closely related to the political intervention. The condition causes two problems, namely: (1) monetary expansion on the income and wealth distribution leads to the increasing gap between the poor and the rich. The increasing gap causes the labor market to be less flexible and increases unemployment, (2) monetary expansion causes less fluctuation but eventually increases unemployment. Using the Uni European Countries data, Pallis (2006) investigated the relationship between inflation and unemployment in the 10 new European Union countries find that in almost all countries the interaction between the price inflation rate and the unemployment level took place in a rather long time period, reaching in some cases the lag of year four. In Pakistan, Ul-Haq et al. (2012) provided further support for the existence of a long-term relationship between unemployment and inflation. On the other hand, the outcomes of VECM revealed a positive and significant correlation between inflation and unemployment either in the long term or the short term.

Other studies demonstrate the two-way causality between inflation and unemployment. For example, Arshad (2014) shows the two-way causality between inflation rate and unemployment in Pakistan, implying that inflation causes unemployment, and unemployment causes inflation. The data suggest that inflation rate explains the variance of unemployment better than economic growth while unemployment contributes to the variance of inflation more than economic growth. In Sri Lanka, Thayaparan (2014) finds the two-way causality between inflation and unemployment, implying that inflation causes unemployment and unemployment causes inflation. The findings indicate that both unemployment and inflation significantly affect the Sri Lankan macroeconomic conditions. Next Bhattarai (2016) finds bidirectional causality as well as cointegrating relationships between unemployment and inflation among the OECD countries. Estimates of a vector autoregression (VAR) model on these trade-offs also support such hypothesis.

Overall, these studies show varying results such as one-way causality, two-way causality, and no causal relationship between inflation and unemployment. Further, these studies also use different analytical models, such as Granger Causality, Johansen Cointegration, Autoregressive Distributive Lag, Error Correction Model, Vector Error Correction Model, Panel Data, Vector Autoregression, and etc. It can be concluded from the previous discussion that there is an uncertain relationship between inflation and unemployment of different economies in the certain period.

2. Research methods

This study uses the secondary data from the central bureau of statistics and the world bank publication. More specifically, the study relied on the time-series data from 1984 to 2017. Further Granger Causality and Vector Autoregression used to analyze the data. Before running the Granger Causality and Vector Autoregression model, this study initially ran the stationary and the lag length test. The following are the models for the stationary test and the test statistic (Brooks 2008).

\[ \Delta Y_t = \varphi Y_{t-1} + \varphi_t; \]  
\[ \text{Test statistic} = \frac{\varphi}{\delta E \varphi}; \]  
\[ X_t = \sum_{i=1}^{m} a_i X_{t-i} + \sum_{j=1}^{n} b_j Y_{t-j} + \mu_t; \]

After running the stationary test, this study ran the lag length test. There are various approaches to select the optimal lag length, such as Likelihood Ratio, Final Prediction Error, Akaike Information Criterion and Schwarz Information Criterion (Rosadi 2012). This study uses the Akaike Information Criterion (AIC). The minimum value of the AIC suggests the optimal lag (Ivanov and Kilian 2005). After completing the lag length test, this study ran the Granger Causality test (Rosadi 2012):
The above equation indicates that $X_t$ is inflation, and $Y_t$ is unemployment, while $\mu_t$ and $V_t$ are error terms that are assumed to contain no serial correlation and $m = n = r = s$. The Granger Causality test produces four possible results as represented by the following equations:

1. If $\Sigma aj \neq 0$ and $\Sigma bj = 0$, then there is a one-way causality from inflation to unemployment.
2. If $\Sigma aj = 0$ and $\Sigma bj \neq 0$, then there is a one-way causality from unemployment to inflation.
3. If $\Sigma aj = 0$ and $\Sigma bj = 0$, then there is no causal relationship between inflation and unemployment.
4. If $\Sigma aj \neq 0$ and $\Sigma bj \neq 0$, then there is a two-way causality between inflation and unemployment.

Further, this study ran the Vector Autoregression (VAR) after completing the Granger Causality test. The Vector Autoregression (VAR) is commonly used for forecasting systems of interrelated time series and for analyzing the dynamic impact of random disturbances on the system of variables. The reduced form VAR approach sidesteps the need for structural modeling by treating systems of interrelated time series and for analyzing the dynamic impact of random disturbances on the system of variables. The reduced form VAR approach is given by the following equation in the Vector Autoregression ($p$) with $k$-endogenous variable $y_t = (y_{t1}, y_{t2}, \ldots, y_{tk})$ (Lütkepohl 2006).

$$y_t = A_1 y_{t-1} + \ldots + A_p y_{t-p} + Cx_t + \epsilon_t,$$  \hspace{1cm} (5)

where:

- $y_t = (y_{t1}, y_{t2}, \ldots, y_{tk})'$ is a $k \times 1$ vector of endogenous variables;
- $x_t = (x_{t1}, x_{t2}, \ldots, x_{td})'$ is a $d \times 1$ vector of exogenous variables;
- $A_1, \ldots, A_p$ are $k \times k$ matrices of lag coefficients to be estimated;
- $C$ is a $d \times k$ matrix of exogenous variable coefficients to be estimated;

- $\epsilon_t = (\epsilon_{1t}, \epsilon_{2t}, \ldots, \epsilon_{kt})'$ is a $k \times 1$ white noise innovation process, with $\mathbb{E}(\epsilon_t) = 0$, $\mathbb{E}(\epsilon_t \epsilon_s') = \Sigma \epsilon$, and $\mathbb{E}(\epsilon_t \epsilon_s') = 0$ for $t \neq s$.

3. Results

Table 1 below shows the results of the stationarity test using the Augmented Dickey-Fuller (ADF) method.

Table 1 indicates that inflation is stationary at the order of integration of level or I(0) while the unemployment level is not stationary at the order of integration of level, prompting us to have the first difference technique. The order 1 or I(1) differencing shows that DUnemployment does not contain the unit root anymore because it is now stationary. Further, determine the optimal length of lag by using Lag Length Test as can be seen in Table 2.

Table 2 suggests the optimal lag to indicate the dependence of a variable on its lagged value and other endogenous variables is lag 3, implying that we have to use lag 3 to investigate the causality between inflation and DUnemployment. This decision is indicated by the Akaike Information Criterion (AIC) value of 9.931972 that is smaller than the AIC values at the other lags. After ran the lag length test, this study ran the Granger Causality test using lag 3. The results of Granger Causality test can be seen in Table 3.

Table 3 reveals that the null hypothesis proposing that DUnemployment does not Granger Cause inflation is rejected, implying that DUnemployment exhibits the Granger Cause on inflation. The results suggest that DUnemployment granger cause inflation, but not vice versa.
versa. The decision of rejecting the null hypothesis is based on the probability value of 2.E-08 that is lower than α = 5%. After ran the Granger Causality, this study ran the Vector Autoregression. Variables in a Vector Autoregression model are determined simultaneously and rely more on historic patterns of data to establish relations between unemployment and inflation than economic theories (Bhattarai 2016). The results of Vector Autoregression can be seen in Table 4.

Table 4 indicates that a simple Vector Autoregression model with three lags on inflation and DUnemployment shows that Inflation is significantly influenced by DUnemployment(–2), DUnemployment(–3) and Inflation(–1). It implies that the influence of DUnemployment(–2), DUnemployment(–3) and Inflation(–1) have a large contribution to the movement of inflation in Indonesia. Estimates of Vector Autoregression also support by the Impulse Response Functions (IRFs). The results of the Impulse Response Functions (IRFs) can be seen in Figure 1.

Impulse Response Functions (IRFs) was calculated for DUnemployment and inflation to address the reaction of the economy to external changes (shocks). The results of the IRFs analysis show that there is a trade-off between inflation and DUnemployment as shown by the IRFs of DUnemployment to inflation. Overall, estimate results of Granger Causality, Vector Autoregression, and Impulse Response Functions (IRFs) prove that the DUnemployment is more instrumental to explain inflation in Indonesia.

Further, this result is supported by Touny (2013) who finds the positive effect of unemployment on inflation. The normalized cointegration equation reveals that unemployment gap has a long-run positive effect on the changes in the inflation rate, which is consistent with “Lucas Critique” where a policy of inflation would fail to reduce the unemployment rate in the long run, because workers would eventually adjust their expectations of inflation. Further, the more rapid the reduction in the unemployment rate, the less disinflation is achieved at each unemployment rate level. Even at the cases where the unemployment rate is very high, the inflation rate falls little and thus the economy is moving too rapidly out of the recession (Pallis 2006). Other findings by Ul-Haq et al. (2012) also suggest that policy makers should pay special attention to this relationship between inflation and unemployment when they are going to design macroeconomic policies.

Thus, when inflation does not support DUnemployment, it is necessary to analyze factors that affect DUnemployment. Table 5 below presents the information on the number of
the population 15 years of age or over who worked by the main employment status.

Table 5 indicates the sharp increase in the number of the working-age population with the employee status, both in absolute and relative terms. The number of employees in 2001 was 26,579,000 or 29.7% of the total working age population. The number increased to 47,420,633 or 38.08% of the total working age population in 2017. Increased investment mainly drives the increasing number of employees. Djambaska and Lozanoska (2015), Yelwa et al. (2015), Touny (2013), Israel (2015), UI-Haq et al. (2012), Bhattarai (2016), and Pallis (2006) support the results by arguing that

![Graphs showing impulse responses to DUnemployment and inflation shocks](image)

**Figure 1. Impulse responses to DUnemployment and inflation shocks**

Table 5. The Population 15 Years of age or over by the main employment status (2001–2017) (source: Badan Pusat Statistik (2017a), processed)

<table>
<thead>
<tr>
<th>Number</th>
<th>Main Employment Status</th>
<th>2001</th>
<th>2017</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Self-employed</td>
<td>17,451,704</td>
<td>21,849,573</td>
</tr>
<tr>
<td>2</td>
<td>Employer Assisted by Temporary/ Unpaid Worker</td>
<td>20,329,073</td>
<td>21,275,899</td>
</tr>
<tr>
<td>3</td>
<td>Employer Assisted by Permanent/ Paid Worker</td>
<td>2,788,878</td>
<td>4,446,024</td>
</tr>
<tr>
<td>4</td>
<td>Employee</td>
<td>26,579,000</td>
<td>47,420,633</td>
</tr>
<tr>
<td>5</td>
<td>Casual Agricultural Worker</td>
<td>3,633,126</td>
<td>5,360,306</td>
</tr>
<tr>
<td>6</td>
<td>Casual Non-Agricultural Worker</td>
<td>2,439,035</td>
<td>6,021,760</td>
</tr>
<tr>
<td>7</td>
<td>Family/ Unpaid Worker</td>
<td>17,586,601</td>
<td>18,164,654</td>
</tr>
<tr>
<td>8</td>
<td>No Answer</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Total</td>
<td></td>
<td>90,807,417</td>
<td>124,538,849</td>
</tr>
</tbody>
</table>
investment is a determining factor in reducing unemployment. Further, another factor that affects the number of unemployment is the industry in which the population work. Table 6 below shows the data on the population 15 years of age or over who worked by industry.

The proportion of the population above 15 years who worked at the primary sectors (agriculture, plantation, forestry, hunting, and fishery) declined sharply. In 1991, 53.29% of the working population worked in the primary sectors, and proportion declined to 31.74% in 2016. The agricultural sector dominates the primary sectors because of most population work in this sector (Thayaparan 2014, Yelwa et al. 2015, Kebschull 1987, Israel 2015). Edelman (2013) confirms the findings by suggesting that farmers in Latin America and Indonesia live in communities with exclusive land rights and most of them use the lands for agricultural activities. The agricultural works in Indonesia are heavily affected by the season factor, especially before 2005 because of the less developed irrigation system. More specifically, the agricultural sector greatly depends on the sufficient availability of rainfall. Farmers begin to cultivate their soils after rain falls. Rain usually starts to fall in October and the dry season starts in April. The following Table 6 displays the open unemployment level based on the February and August surveys. February is in the rainy season while August is in the dry season. The open unemployment rate was higher in August, a month in the dry season than in February, a rainy month. The average August unemployment rate was 8,599,944 while in February the average unemployment level was 8,599,676. The proportion of the population working in the agricultural sector was so high that the season factor significantly affected the unemployment rate. However, the annual difference of the unemployment level between these two months tended to decline. There was even no difference of the open unemployment level between February and August in 2017 (Badan Pusat Statistik 2017d). Further estimates of an Independent Sample T Test also support the data. Estimate results can be seen in Table 7.

The significance value (2 tailed) of 1.000 is bigger than the tolerance value of 5% (0.05) implies that there is no difference between February and August unemployment rate. The more developed irrigation system reduce the farmers’ dependence on rainfall and eventually on the season. Next, the Indonesian working-age population who attended

Table 6. Population 15 years of age or over who worked by main industry (1991–2016) (source: Badan Pusat Statistik (2017b), processed))

<table>
<thead>
<tr>
<th>Number</th>
<th>Main Industry</th>
<th>1991</th>
<th></th>
<th>2016</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Agriculture, Plantation, Forestry, Hunting, and Fisheries</td>
<td>39,385,946</td>
<td>53.29</td>
<td>38,291,111</td>
<td>31.74</td>
</tr>
<tr>
<td>2</td>
<td>Mining and Quarrying</td>
<td>551,581</td>
<td>0.75</td>
<td>1,311,834</td>
<td>1.09</td>
</tr>
<tr>
<td>3</td>
<td>Manufacturing Industry</td>
<td>7,712,468</td>
<td>1.43</td>
<td>15,975,086</td>
<td>1.24</td>
</tr>
<tr>
<td>4</td>
<td>Electricity, Gas, and Water</td>
<td>148,480</td>
<td>0.20</td>
<td>403,824</td>
<td>0.33</td>
</tr>
<tr>
<td>5</td>
<td>Construction</td>
<td>2,415,002</td>
<td>3.27</td>
<td>7,707,297</td>
<td>6.39</td>
</tr>
<tr>
<td>6</td>
<td>Trade, Restaurants, and Accomodation Services</td>
<td>11,190,391</td>
<td>1.14</td>
<td>28,495,436</td>
<td>23.62</td>
</tr>
<tr>
<td>7</td>
<td>Transportation, Warehousing, and Communication</td>
<td>2,475,803</td>
<td>3.35</td>
<td>5,192,491</td>
<td>4.30</td>
</tr>
<tr>
<td>8</td>
<td>Financial, Real Estate, and Business Services</td>
<td>515,401</td>
<td>0.70</td>
<td>3,481,598</td>
<td>2.89</td>
</tr>
<tr>
<td>9</td>
<td>Community, Social, and Personal Services</td>
<td>9,377,036</td>
<td>12.69</td>
<td>19,789,020</td>
<td>1.40</td>
</tr>
<tr>
<td>10</td>
<td>Undefined</td>
<td>139,516</td>
<td>0.19</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Total</td>
<td></td>
<td>73,911,624</td>
<td>100.00</td>
<td>120,647,697</td>
<td>100.00</td>
</tr>
</tbody>
</table>

Table 7. Independent Sample T Test for Unemployment on February and August (1986–2017)

<table>
<thead>
<tr>
<th>Levene's Test for Equality of Variances</th>
<th>t-test for Equality of Means</th>
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<tr>
<td></td>
<td></td>
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<tr>
<td>Unmpl</td>
<td>.002</td>
</tr>
<tr>
<td>Equal variances assumed</td>
<td></td>
</tr>
<tr>
<td>Equal variances not assumed</td>
<td></td>
</tr>
</tbody>
</table>
school or performed the housekeeping increased. The number of economically inactive women due to housekeeping increased both in absolute and relative terms (Ehrenberg and Smith 2012). In 2005, the number of the working age population who performed the housekeeping (mostly women) was 17,275,478 or 17.08% of the total labor force. The number increased to 36,078,772 or 18.78% of the labor force in 2017 (Badan Pusat Statistik 2017c). The estimates of an Independent Sample T Test also support the data. Estimate results can be seen in Table 8.

The significance value (2 tailed) of 0.000 is lower than the tolerance value of 5% (0.05) implies that there is a difference between the Indonesian working-age population who attended school and performed the housekeeping. Ehrenberg and Smith (2012) confirm the data by arguing that women spend a significant portion of their time to housekeeping such as cooking or taking care of their children. Women prefer housekeeping to enter the labor market because housekeeping is also a productive activity.

Further, the number of the population 15 years of age or over who attended school increased from 9,147,830 in 2005 to 15,244,852 in 2017. However, the proportion of the working-age population who attended school decreased from 9.04% in 2005 to 7.94% in 2017. Higher school attendance decreases the number of unemployment. Hubacek et al. (2007) support the findings by demonstrating that the economic success of the developing Asian countries enhances the quality of life of their population. Most Asian population experience the transition from poverty to sufficient fulfillment of food and clothes. Further, they aspire to not only meet their basic needs of food and clothes, but also to enjoy a higher quality of life from highly nutritious food, life comfort, medical treatments, and other highly qualified services.

**Conclusions**

This study suggests the one-way relationship between inflation and unemployment. More specifically, the Granger Causality, Vector Autoregression, and Impulse Response Functions (IRFs) model show that from 1984 to 2017, DUnemployment causes inflation, but not vice versa. The results imply that the Phillips model (Phillips 1958) that proposes the reciprocal relationship between inflation and unemployment is not empirically supported in Indonesia. These findings also different with the most recently by Blanchard (2016) who find that The US Phillips curve is alive and well (or at least as well as it has been in the past).

Various factors affect the Indonesian unemployment rate, such as: (1) The season factor significantly affects unemployment, albeit with the declining magnitude, because the agricultural sector still absorbs a significant portion of the Indonesian labor force; (2) Increased income encourages young labor force (15–19 years) to delay entering the labor market but to continue their studies; and (3) Better economic condition also increases the number of non-labor force. More specifically, women prefer becoming housewives (caring for their households) in entering the labor market because caring for households is also a productive activity (Ehrenberg and Smith 2012).

Inflation is a less effective policy instrument to overcome the unemployment problem in Indonesia. This argument implies that increasing the inflation rate is ineffective to reduce the unemployment rate. Numerous facts indicate that other variables affect the Indonesian unemployment rate. However, it is viable to increase the unemployment rate to control inflation, although this policy has to be implemented carefully. Further, the Indonesian geographical condition that consists of thousands of islands likely causes the implementation of macro policies to take a longer time because of the greater needs to adjust for the inter-region differences. Thus, the use of the panel data model likely accounts for the possible inter-region variances better.

**References**


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Table 8. Independent Sample T Test for Working Age Between Attend School and Housekeeping (1986–2017)

<table>
<thead>
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<th>Levene's Test for Equality of Variances</th>
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G. Sasongko, A. D. Huruta. The causality between inflation and unemployment: the Indonesian evidence


