

The p. 1) to  
investigate the  
gender gap



The gender wage gap is defined as “ the difference between the median earnings of men and women relative to the median earnings of men” 1.” The median wage of a woman working full-time is 85% that of a man” in rich and middle-income countries that the OECD consists of 2. This review will concentrate on five studies that all examine the existence of the gender wage gaps in many countries, involving several variables that are tested, and different methods used to gather the data.

The inspiration from Chuang, Lin, and Chui (2017, p. 1) to investigate the gender gap in the financial industry derived from the part that industry plays in affecting gender wage gaps for export-oriented countries. Further, they state as Taiwan is a well known export-oriented economy, the analysis could serve as a representative case study. The sample chosen was taken from individual level data from the “ Manpower Utilization Survey” from the years 1978 to 2013 which each year consists of approximately 18000 randomly drawn sample households. The sample used in the analysis is restricted to paid employees in the private sector.

Through the decomposition, examination, and the breakdown of the overall gender wage gap, they examine the interindustry gender wage gap in Taiwan, particularly focusing on the financial industry. They found that during the period sampled, 2-14% of the overall gender wage gap can relate to workers’ industry association. Chuang, H-L, et al., (2017, p. 4) state that they will use Mincer’s human-capital earnings function as a theory underlying the wage equation specification and use it to compute the log-wage for a representative male and female worker in industry j. Using the Oaxaca and Blinder (OB) strategy Chuang, H-L, et al., (2017, p. 5) decompose the gender

wage gap in industry  $j$  into the following explained and unexplained components.

The first four terms on the right-hand side of the model display the unexplained components and the last two terms represent to the explained wage gap in the industry. However, the OB decomposition suffers from the index-problem that can be split into two identification problems (IP1 and IP2). Fields and Wolff defined the industry wage gap for industry  $j$  as: The first term is present in the equation in order to capture the effect of choosing the reference group but does not get rid of IP2. Horrace and Oaxaca suggested four alternatives to overcome IP2 as follows;  $\cdot \cdot \cdot$  Both are free from identification problem 2 but still are affected by identification problem 1. However, the two ranking measures, need to imitate the critical values to be able to perform a statistical inference. Consequently, a measure that was developed by Lin (2007b, 2010) that resolves both IP1 and IP2 is implemented. This further makes available a standard error for the significance test. is recognised as it is free from the choice of the non-discriminatory wage structure that is not observable as well as the left-out reference group of any of the dummy variables.

Chuang, H-L, et al., (2017, p. 6) report that industry variables are of the most interest in studying the gender wage gap, with the mining industry as the reference group. In order to assess the contribution of the industry dummies towards explaining the gender wage gap the OB decomposition techniques are used. From the results it can be seen that a larger fraction of the explained component is due to the female-based calculation. This explained proportion rises when industry dummies are present by 4-10% at 2-8% for

the female-based calculation and 5-14% when it comes to the male-based calculation highlighting the importance of the including industry variables. The financial industry shows the smallest gender wage gap with -0.

0483 based on and -0. 0494 based on while the biggest gap is portrayed through the service industry at -0. 1659 based on . Previously stated in the paper, if an industry has a ranking that is high, which means the ranking number is small, then the wage gap for that particular industry is small. Ranking 8th place based on both and from 1978 to 1991 (except for 1988), females faced the largest gender wage gap in the mining industry. Year after year different industries rank highest before 1997, but the mining industry although it does not rank lowest has the lowest rankings since 1991. In 1978 agriculture is the most beneficial industry for women based on the - group measures however from 1979 to 1997 excluding 1992 and 1995 it changes to the construction industry. Conversely, the -group measures the trading industry as the highest-ranking industry in 1978 and after 1996 the financial industry.

From 1998 excluding the year 2012, both groups signal that the highest-ranking industry is the financial industry. When the ratio of female employment across industries is studied further, after the increase in female employment over time, almost all sectors have women employed in them. In the mining industry, between the period 1978-1991 and 1992-2013, average proportion of female employment rises by 1.

83% and 4. 12% in the financial industry. On the whole, the difference in wages for females and males in the financial industry is very little.

Overall evidence shows that Femaleworkers overtime gain the most beneficial wages from the financial industry. L. N. Christofideset al (2013) focus their research on understanding the gender wage gap across26 European countries by using data from the European union statistics onIncome and Living conditions (EU-SILC) in 2007. The two samples that areexamined are the “ working sample” and the “ alternative sample”.

The workingsample consists of workers that are between the ages of 25 and 54 who are notstudents, handicapped, retired, doing compulsory community or military service, or have given up a business. The alternative sample that is also known as theFTFY sample comprises of workers that must have worked full-time for the wholeof the previous year as well as the requirements of the working sample. Ordinaryleast squares (OLS) is used by L. N. Christofides et al (2013 p. 89) to estimatean hourly earnings equation by gender which includes the characteristics thatare relevant and available from the EU-SILC data. Results obtained from OLSshow that the actual gender wage gap is equal to the predicted total gap and generally, the selection-adjusted gap is even larger indicating positive selection is atwork. Secondly, the average difference in male andfemale earnings is decomposed following Oaxacaand Ransom (1994) as follows,  $M - F = (M - F)N + M(M - N) + F(N - F)$  The first term  $((M - F)N)$  measuresthe explained component, the second  $(M(M - N))$ the male advantage and the third  $(F(N - F))$  the femaledisadvantage.

The addition of both the male advantage and female advantage representsthe unexplained part. For a number of countries, the unexplained part of thetotal is found to be larger than the explained component implying

that there may be an existence of female disadvantage and the data that is accessible does not explain the behaviour of earnings. The portion that is explained is negative in Belgium, Greece, Hungary, Iceland, Italy, Luxembourg, Poland, Portugal, Slovenia, and Spain proposing that female characteristics are greater than that of males. Furthermore, in the majority of countries, the offered and total wage gaps are smaller in the working sample of part-year and part-time workers compared to that of the FTFY sample. In ten countries of the 26, the public sector of the working sample has a larger female disadvantage and in the case of the private sector, eight countries have a smaller gender wage gap. In the alternative sample, nine countries have a reduced gender gap in the public sector and is ten when there is a larger disadvantage. These results are only slightly coherent with the understanding that where FTFY jobs are involved, the public sector is more progressive (N.

L. Christofides et al., 2013, p. 92). As the analysis of possible “sticky floors” and “glass ceiling” effects are not allowed for by the decomposition of the mean variations, Melly (2005) uses a method called the quantile regression methodology that decomposes along quantiles of the wage distribution that addressed any selection issues that may occur. This method lets the characteristics of workers at different points of the wage distribution have different effects and in turn affect the decompositions at each point. When comparing the mean values in the Oaxaca and Ransom decompositions with the quantile regression total and unexplained gaps at the 50th percentile, more countries have unexplained components that surpass the total wage

gaps. However, evidence from Austria, Estonia and the UK show that the total exceeds the unexplained gap for all the quantiles.

Hence, the quantile results highlight the conclusion that a considerable portion of the earnings gap continues to be unexplained. Evidence of sticky floors in twelve countries is present with the strongest results from Cyprus, France, Italy, Luxembourg, Slovenia, and Sweden. Prominent glass ceiling effects are demonstrated in eleven countries that are Denmark, Germany, Hungary, the Netherlands, Norway, the Slovak Republic, the Czech Republic, Finland, Iceland, Slovenia, and the UK. In the FTFY sample, stick floor behaviour is present in twelve countries as well but instead of Belgium, Spain is added. Countries exhibiting glass ceiling effects conversely have increased to twenty-one instead of eleven. Cyprus, Estonia, Lithuania, Portugal, and Spain do not show evidence of glass ceiling behaviour. The prevalence of this behaviour in the FTFY sample is coherent with the interpretation that women are more likely to be at a disadvantage in FTFY positions, specifically when they are high-paying ones. Generally, in the public sector, female employees have a lower disadvantage than in the economy in eight countries and are at a higher disadvantage in ten countries and in nine countries a lower gender disadvantage is present compared with the private sector.

All findings indicate that gender gaps are bigger when individuals must be in full time and full year employment. D. Antonczyk et al. (2010, pp. 835-847) concentrate their study on exploring the link between the recent rise in wage inequality between 2001 and 2006 in West Germany, as well as the fall in collective wage bargaining and the progression of the gender wage gap for

West Germany. The focus of this review will be more on wage inequality and the gender wage gap.

The sample involves repeated cross-sections of the earnings of 440, 000 employees, between the ages of 25 and 55, in 17, 000 establishments in 2001 and 750, 000 employees in 22, 600 establishments in 2006 taken from the employer-employee data set. All employees are full-time workers. There are significant changes in the wage distribution over the years, for example, for both males and females' real hourly wages fall below the median whereas they rise for the quantiles that are beyond the median. Overall, this leads to an growth in wage distribution. Women are able to achieve most comparatively to men in the bottom part of the wage distribution from 2001 to 2006. D. Antonczyk et al.

(2010, p. 840-842) propose a sequential decomposition from both the cross-sections of data in 2001 and 2006. Suggested in DiNardo et al. (1996) and developed further in Chernozhukov et al. (2008) and Antonczyk et al. (2009), the decomposition aims to capture wage structure that may be influenced by wage bargaining, firm characteristics, and personal characteristics. The decomposition is split into coefficient effects (personal, firm and bargain coverage) the residual change in overall wage level overtime and characteristic effects (bargain coverage, firm and personal).

The most significant component that increases the wage inequality is shown by the changes in the firm coefficients. In addition, changes in residual wages and personal coefficients lead to the rise in wage inequality whilst personal characteristics work against this trend. Even though these personal



characteristics add to the inequality of wages mainly for females at the bottom, the effect is typically irrelevant. Changes of wage differences within and between industry mainly push inequality up which could imply that the changes of firm wage policies may have increased within and between industries, perhaps due to the more extensive use of irregular payments schemes. Wage inequality for both females and males could result from the unexplained time trends with wages at the top of the distribution escalating and wage at the bottom declining. This trend causes a consistent reduction in wages for females of approximately 1.3 percentage points, which is however, quite insignificant. Overall, all workplace related effects add to the strong increase in wage inequality.

S. Machin, P. A. Puhani (2003) assess the relevance of subject of degree in explaining a sizeable proportion of the gender wage differential amongst graduates. With data taken from the labour force surveys of both the UK and Germany in 1996, they estimate separate log (earnings) functions for men and women graduates that do and do not control for subject degree. The log wage difference is slightly smaller in the UK than it is in Germany at 0.208 and 0.280 respectively. This could coincide with the view that in Germany, women are not as advanced in the wage hierarchy compared to Britain. However, another reason for this could be that the data collected from the German labour force survey only measures net income whereas in the UK the data consists of gross wages. Decomposition of the earnings is used (S.

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Machin, P. A. Puhani 2003, p. 396) in three specifications. The first specification controls for age and age squared like the typical Mincerian wage equation. Specification two adds a number of other components that are expected to influence wages, specifically, industry, region, and dummies for part-time and public-sector employment.

Finally, the third specification includes occupation. Although, there is potential endogeneity meaning it could correlate with the error term. From specification one, the results produce analogous results for both countries with approximately 21%-24% of the gender wage gap between graduates explained by age.

The explained gap almost more than doubles when the subject of degree is added for both countries. Additionally, by controlling for less detailed degree categories the explained component of the gap increases from 24% to 43% for the UK and from 21% to 36% in Germany. However, if detailed degree types are controlled for then the explained gap further increases to 56% and 41% in the UK and Germany respectively. This explained increase is statistically significant in explaining the differentials of wage between females and males. When the same analysis is undertaken with specification two which incorporates more variables, a moderately large effect of controlling for subject degree is present. In both countries, combined subject of degree categories account for a 2% wage premium, doubling to 4% in the UK.

However, for this model there is not much extra impact from detailed degree subject in Germany. The percentage point increase of the gap explained by

degree type is 7 and 16 for Germany and UK respectively in specification 3 showing that even with occupation, subject degree still matters. On the whole, the results for both countries are very similar and shows the importance of subject degree in explaining the wage gap differences between male and female graduates. S. Brown et al. (2011) study data from the British household panel survey between 1991 and 2008 of over 5000 private households to examine whether there is a gender wage gap and if so does the presence of children play an influential role in determining the gap.

The sample consists of 12,921 observations with 53% of the sample gathered being female. S. Brown et al. (2011, p. 89) decompose the reservation wage gap into an equation with five different samples used to investigate the effect of children on the gender reservation wage gap. The results show that the reservation wage gap is statistically significant and positive and 78% of the difference in wages continues unexplained.

From the explained component, the number of children is the most important factor. Children alone can explain why women have higher reservation wages than men, as the negative coefficient on the variable proposes that it reduces the gap between reservation wages between men and women. Without the presence of children, the unexplained part of the gap rises to 99% portraying, observed discrimination for those with children may have a large influence here. A variable that leads to the reservation wage gap widening is education.

The results indicate that there is an existence of a reservation wage gap between males and females and the presence of children plays a substantial

role in determining this gap. Without children almost none of the gap is explained.