Holmberg, Pär; Newbery, David

Working Paper
The Supply Function Equilibrium and Its Policy Implications for Wholesale Electricity Auctions

IFN Working Paper, No. 812

Provided in Cooperation with:
Research Institute of Industrial Economics (IFN), Stockholm

Suggested Citation: Holmberg, Pär; Newbery, David (2009) : The Supply Function Equilibrium and Its Policy Implications for Wholesale Electricity Auctions, IFN Working Paper, No. 812, Research Institute of Industrial Economics (IFN), Stockholm

This Version is available at:
http://hdl.handle.net/10419/81375

Terms of use:
Documents in EconStor may be saved and copied for your personal and scholarly purposes.
You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.
IFN Working Paper No. 812, 2009

The Supply Function Equilibrium and Its Policy Implications for Wholesale Electricity Auctions

Pär Holmberg and David Newbery
The supply function equilibrium and its policy implications for wholesale electricity auctions*

Pär Holmberg † and David Newbery ‡

November 5, 2009

Abstract

The supply function equilibrium provides a game-theoretic model of strategic bidding in oligopolistic wholesale electricity auctions. This paper presents an intuitive account of current understanding and shows how welfare losses depend on the number of firms in the market and their asymmetry. Previous results and general recommendations for divisible-good/multi-unit auctions provides guidance on the design of the auction format, setting the reservation price, the rationing rule, and restrictions on the offer curves in wholesale electricity auctions.

Key words: Wholesale electricity markets, supply function equilibria, competition policy, market regulation

JEL Classification D43, D44, C62, L94

*Research support by the ESRC to the Electricity Policy Research Group, EPRG, under the programme Towards a Sustainable Energy Economy is gratefully acknowledged. Pär Holmberg has been financially supported by the Research Program The Economics of Electricity Markets. We are indebted to Andrew Philpott for discussions on the unique SFE for elastic demand. We are also very grateful for comments to this manuscript from Lawrence Ausubel and other participants at the workshop "Designing electricity auctions" in Stockholm, Sept. 2009.

†Research Institute of Industrial Economics (IFN), Stockholm. Visitor to the EPRG.

‡Faculty of Economics, University of Cambridge, Research Director, EPRG.
1 INTRODUCTION

In contrast to most other markets, the way price is determined is very well defined in wholesale electricity markets. Each producer submits an offer curve that specifies how much it is willing to produce at different prices. Similarly, consumers and retailers (suppliers), who represent small consumers, submit demand curves specifying how much electricity they want to buy at different prices. Market regulators can influence the price formation and how competitive the market will be by choice of auction format, the level of any price cap, the rationing rule, and by making restrictions on the offer curves. When making their choice regulators should consider the impact on participants contracting and investment incentives under various market designs and rules. Competition authorities also need to predict electricity prices under various counterfactuals – what might happen if a merger or acquisition is accepted? Often authorities are content with using concentration measures, such as the Herfindahl-Hirschman index (HHI), to assess the degree of competition in the market. However, these measures work poorly for electricity markets, where demand and supply must be instantly balanced and where the tightness of reserve margins and transmission capacity constraints can vary considerably over short periods with significant impacts on prices (Borenstein et al., 1999). Thus given installed production capacities, it depends very much on the level and location of instantaneous demand whether the market will suffer from the exercise of market power.

This paper qualitatively assesses the two leading auction formats, the uniform-price and pay-as-bid formats, and other rules of electricity auctions using supply function equilibria under uncertain demand. We provide new results highlighting how short-run welfare losses depend on the number of firms in the market and their asymmetry.

The supply function equilibrium (SFE) model was originally developed by Klemperer and Meyer (1989) and first applied to the electricity market by Green and Newbery (1992) and Bolle (1992). It is a game-theoretic model of competition in wholesale electricity markets. It assumes that each producer chooses its offer curve in order to maximize its profit, given demand and offer curves chosen by competitors. In concentrated markets, producers may be able to tacitly collude to higher electricity prices, as Sweeting (2007) suggests happened in Britain, but this effect is only briefly considered below. The SFE is a static equilibrium, i.e. it is assumed that all producers will play their profit-maximizing strategies repeatedly, irrespective of what happened in previous periods. In practice it may take a while before each producer has figured out its best offer strategy given the residual demand it faces. This learning is facilitated in markets that disclose individual or aggregated offer curves to the auction, such as the Amsterdam Power Exchange (APX), illustrated in figure 1. This figure shows the determination of the market clearing price (MCP) for hour 12 on 26 June 2007, illustrating a part of the ladders of offers and
bids and showing that 1,942.4 MW was traded at a MCP of 58.83 Euros/MWh.\footnote{Typical capacity connected to the Dutch system would be over 15,000 MW and the APX covers a wider area than just the Netherlands, so a relatively small fraction of capacity is traded on the APX.}

1.1 The wholesale electricity market

Electricity is produced by many different technologies that often have very different marginal costs. The production cost of a plant is primarily determined by fuel costs and its efficiency that are well-known and common knowledge. The plants of a producer are used in merit order, starting with the lowest marginal cost, such as nuclear power or hydro-power. Last in the merit-order are peaking plants, such as open-cycle gas turbines burning natural gas or oil with high variable and low capital costs. The merit-order implies that producers’ marginal costs increase with output. There are some local deviations from this trend, as start-up costs introduce local non-convexities, but these are normally neglected in market analyses (though not in optimal scheduling programmes). Although electricity may be produced by various technologies, it is still a completely homogeneous good suitable for trading on commodity exchanges and auctions.

In wholesale electricity markets, producers sell electricity to retailers. In their turn, retailers sell electricity to consumers in the retail market. Electricity consumption is to a large extent determined exogenously, e.g. by the weather and work-days or holidays, and is very inelastic,
especially close to the time of consumption. This limited flexibility means that retailers’ market power is small compared with that of generators, which can be significant.

Due to restrictions on the rate that fossil generation can ramp-up output, particularly from a cold start, production plans are scheduled the day before delivery, and the day-ahead (or prompt or spot) market is an important component in this planning process. A well-designed liquid market can provide the strike prices for financial contracts. The day-ahead market is organized as a double auction to which consumers and producers submit non-increasing bid curves and non-decreasing offer curves, respectively, as shown in figure 1. The market clearing price (MCP in figure 1) is determined by the intersection of the bids and offers. There is normally a separate price and auction for each delivery period, typically a half-hour or hour but which can be as short as 5-15 minutes.

Electricity is special in that supply must equal consumption at every instant, because it is very expensive to store electrical energy on a large scale. The system operator uses a real-time or balancing market to make necessary adjustments to production (and consumption to the extent that it bids into the market) during the delivery period by accepting additional power production from, or by selling back electricity to, producers. Offers to the balancing market are submitted before the delivery period starts and are valid during the whole delivery period, but demand is uncertain when offers are submitted. System imbalances arise because of unexpected changes in wind and temperature, unexpected production outages or unexpected transmission-line failures. These are normally handled by holding scheduled reserve capacity, measured as available production capacity less expected demand.

In very extreme cases, this scheduled reserve capacity may be insufficient to meet the system imbalance caused by multiple unexpected events, and an outage or loss of load occurs. The loss-of-load probability (LOLP) is typically very small, but always positive. No matter how large the reserve margin, sufficiently many simultaneous unexpected events that decrease the production or reserve capacity, or increase demand, will lead to a power shortage. LOLP during any delivery period can be estimated ex ante from the scheduled reserve margin, the probability distribution of demand and from the probabilities of having production failures in individual plants. Newbery (1998b) shows that the LOLP estimated ex ante by the system operator in Britain decreases exponentially with the scheduled reserve margin. Using data provided in Newbery’s paper, and adjusting for the system operator’s consistent overestimation of LOLP, we roughly estimate LOLP during an half-hour to 0.1% for a scheduled reserve margin of 10% for the British market from 1990-95, and that it roughly decreased by a factor 100 for every additional 10% of reserve capacity.\(^2\) However, these estimates are very uncertain, and only intended to give some feeling

\(^2\)The values are estimated from Fig. 5 in Newbery (1998b) and that capacity payments in the old pool were
for the magnitude of LOLP and its relation to the reserve margin. Normally, the variance of the demand distribution is small and predictable, with most uncertainty lying on the supply side.

In the rare situations when electricity demand exceeds market supply, demand has to be rationed to avoid a system collapse, and this occurs when the price reaches the price cap. One motivation for the use of price caps is that consumers who do not switch off their equipment when the electricity price becomes very high do not necessarily have a high marginal benefit of power. It may be that the residential consumer is not at home or not aware of the high price. Moreover, residential consumers may not face the real-time price, as their consumption is normally measured over long intervals, typically a year, and often distribution companies do not observe who consumes what in real-time. Such market or information imperfections make it welfare maximizing to ration demand at some very high reservation price, often set equal to the value of lost load (VOLL) (Stoft, 2002). In the English Electricity Pool, VOLL was set at £2,500/MWh in 1990 (5,000 euros/MWh at 2009 prices and exchange rates). These market/information imperfections may be reduced by installing real-time meters (at least if consumers respond) and could be completely removed if the meters can automatically control desired household consumption.

Most electricity wholesale and real-time (or balancing) markets are organized as uniform-price auctions. Hence, all accepted bids from retailers pay the market clearing price and all accepted offers from producers are paid the same price. The British balancing market (strictly, the Balancing Mechanism) is an exception. In this market, all accepted bids pay or are paid their bid. This discriminatory format has also been seriously considered in other electricity markets, e.g. California (Kuhn et al., 2001) and recently also in Italy. In several, mainly European, markets, production is adjusted after market-clearing to ensure that transmission constraints are not violated within any price zone. This is called counter-trading and offers accepted in this post-clearing process are paid their offer price, not the market clearing price (MCP) of that price zone, according to a producer’s individual supply function to the balancing market. Thus proportional to LOLP. It is assumed that the system operator has overestimated the LOLP by a factor 50. This very uncertain factor is based on the reasoning in the appendix of Newbery (1998b).

3The Italian Law Decree n. 185/2008, better known as the "Anti-Crisis Decree" was converted into law by January 27, 2009. As a part of the decree, the Ministero per lo Sviluppo Economico in consultation with the Italian Authority for electricity and gas ("AEEG"), have been instructed to change the trading mechanism on the Day-Ahead Market (or “MGP”) from the current system based on a marginal price rule to a new system based on a pay as bid rule. The change is intended to decrease the average price of electricity traded on the “MGP”.

4Plant that is constrained-off, that is requested not to supply because of excessive local production, is typically paid its lost profit, equal to the MCP less its offer price. Effectively, the producer sells accepted power production at MCP when the market is cleared but if it turns out that his production is infeasible, he has to buy this power back at its offer price in the post-clearing adjustment.
many balancing markets are actually a blend of the uniform-price and the pay-as-bid format. Treasury auctions are similar to electricity auctions in that both are divisible-good/multi-unit auctions and bidders commit to a bid/offer curve in both auctions. Unlike electricity markets, the pay-as-bid format is used in most treasury auctions. Bartolini and Cottarelli (1997) find that 39 out of 42 countries surveyed by them use discriminatory auctions. But it should be noted that the U.S. treasury gradually switched from a pay-as-bid to a uniform-price format during 1992-1998 (Ausubel and Cramton, 2002).

Electricity prices are volatile because electricity is not suitable for large-scale storage, short-run demand is very inelastic and short-run supply can also be inelastic (and in any case varies substantially depending on the fuel and efficiency of the marginal plant). To hedge their risks, market participants can buy and sell various derivative contracts, e.g. futures and forward contracts. These contracts commit the parties to buy and sell the contracted quantity in the spot or real-time market at the agreed delivery price. In most cases the contracts are financial, so no physical transaction takes place. Forward contracts are normally traded over the counter, and futures contracts on power exchanges.

2 The supply function equilibrium

The setting of the supply function equilibrium, originally introduced by Klemperer and Meyer (1989), assumes that production costs are common knowledge and that demand is uncertain. This is reasonable in electricity markets, where technology characteristics and fuel prices are transparent and producers make offers before demand has been realised. They further assume that the shock $\varepsilon$ is additive to the demand schedule, so that $D(p, \varepsilon) = D(p) + \varepsilon$, and production uncertainties are neglected for strategic producers. This considerably simplifies the determination of the residual demand schedule of a producer $i$, for which shocks will again be additively separable, $R_i = R_i(p) + \varepsilon$. These assumptions make the supply function equilibrium model tractable in comparison with multi-unit/divisible good auction models with common or affiliated uncertain values/costs by Wilson (1979) and Ausubel and Cramton (2002), which are often used to analyze treasury auctions. For example, supply function equilibria can be determined analytically for cases with constant marginal costs (Newbery, 1998; Holmberg, 2007) and linear marginal costs with linear demand (Klemperer and Meyer, 1989; Green, 1996; Baldick et al., 2004; Newbery, 2008b).\(^5\) Closed form solutions are also available for symmetric firms and perfectly inelastic demand (Rudkevich et al., 1998; Anderson and Philpott, 2002a). Numerical algorithms make it possible to calculate SFE of markets with asymmetric firms and general cost

\(^5\)There are linear SFE for the case of linear marginal costs, and analytic non-linear solutions have been derived by Newbery (2008).
functions (Anderson and Hu, 2008a; Aromí, 2007; Edin, 2007; Holmberg, 2009a).

The equations originally derived by Klemperer and Meyer (1989) can be explained in the following intuitive way. Even if the demand shock is unknown \textit{ex ante} a producer can still construct an offer curve that leads to an \textit{ex post} optimal profit for each shock outcome. In a pure-strategy Nash Equilibrium (NE) of the game, competitors’ offer strategies are known, so that for each shock outcome a producer knows its residual demand. Thus it can calculate the optimal mark-up for each outcome by applying the monopoly mark-up rule or Ramsey pricing to the elasticity of its residual demand, $\gamma_i^{\text{res}}$ (Tirole, 1988):

$$
\frac{p - C_i'(q_i(p))}{p} = -\frac{1}{\gamma_i^{\text{res}}}. 
$$

(1)

The elasticity of residual demand is straightforward to calculate as long as there are no binding transmission constraints in the power system. With binding constraints, changing an offer in one node will influence the power flows in the system in a non-trivial way, but it is still possible with suitable software to derive the transmission-constrained elasticity of residual demand (Xu and Baldick, 2007). Figure 2 shows that the optimal output for each demand outcome is chosen so that the marginal revenue (MR) equals the marginal cost (MC). This gives a point in the producer’s optimal offer curve. By repeating the procedure for each shock outcome $\varepsilon$, the optimal offer curve can be derived.

Each producer calculates its offer curve in a similar way and the system is in equilibrium when each producer is satisfied that its offer schedule is optimal given the behaviour of all other suppliers. Equilibrium is then calculated from a system of equations as in (1). The equations are differential equations as the optimal mark-up of one producer depends on the slope of competitor’s offer curves. The second-order condition for profit maximizing offer curves in uniform-price auctions is that the marginal cost should increase faster than the marginal revenue at each extremum point, which is the case in figure 2. Holmberg et al. (2008) show that the second-order condition is satisfied for all sets of increasing offer curves satisfying (1) if demand is weakly concave (which includes linear demand).

Without an initial condition or end-condition there is generally a continuum of equilibria bounded by the Bertrand and Cournot equilibria. Figure 3 shows the range of possible equilibria estimated by Green and Newbery (1992) for two possible market structures in the English Electricity Pool. The Cournot line is the optimal offer if all other producers offer a fixed supply. The large range of equilibria can be explained in the following way. Assume for simplicity that marginal costs are constant. Now if competitors’ choose to play Bertrand strategies, i.e. bids are perfectly elastic, then the residual demand of producer becomes perfectly elastic and the best response is, according to (1), to have zero mark-ups, so that the best bid is a Bertrand bid. If competitors’ offers are less elastic then the best response will have increasing mark-ups, so the
Figure 2: The ex post optimal offer is the best response to the residual demands associated with each state of the world.

![Diagram showing supply function and demand variation]

Figure 3: Estimated SFE for England and Wales, 1990.
optimal offer is also less elastic. The range of equilibria is often constrained by capacity (Green and Newbery, 1992; Baldick and Hogan, 2002; Anderson and Hu, 2008a). Genc and Reynolds (2004) analyse in detail how pivotal producers reduce the range of equilibria. In particular, a unique equilibrium will be singled out if maximum demand is high enough to make the capacity constraints of all (but possibly one) firms bind with a positive probability, which could be arbitrarily small (Aromí, 2007; Holmberg, 2007; Holmberg, 2008a). If demand is sufficiently inelastic, the unique equilibrium is selected by the price cap and the capacity constraint, as illustrated in figure 4. We see that the market price is near marginal cost for low demand outcomes and near the price cap for high demand outcomes. Provided that the demand variation is small, this is consistent with results for the auction model by von der Fehr and Harbord (1993), where each producer chooses one price for its whole capacity.

To avoid discontinuities in its optimal offer curve, each producer needs to face a smooth residual demand curve. If demand is sufficiently elastic so that the price cap never binds for the optimal bid curves, then the unique symmetric equilibrium for a positive loss of load probability (LOLP) is given by the symmetric solution where the slope of the supply curve becomes vertical.

Figure 4: For sufficiently large demand variation there is a unique equilibrium that is determined by the price cap and capacity constraints. More producers in the market make bidding more competitive.
For sufficiently large demand variation and sufficiently elastic demand the unique equilibrium is determined by that the offer curve becomes vertical at the capacity constraint. This ensures that all producers have a smooth residual demand at the capacity constraint. This corresponds to the upper supply schedule in Green and Newbery (1992). Note that maximum demand schedule need not cross the smooth offer curve implied by the first–order condition in (1), because at the capacity constraint this curve can be extended vertically up to the price cap without any kinks, as shown in figure 5.

### 2.1 Generalisations of the supply function equilibrium

The supply function equilibrium model developed by Klemperer and Meyer (1989) relied on *ex post* optimality, i.e. a producer could choose its optimal supply curve such that its profit was maximized for each shock outcome. This may not always be possible, if producers choose the best expected outcome before knowing the shock, but would like to have chosen a different offer curve once they know the actual shock. Thus Wilson (2008) analyses supply function equilibria in a transmission network with multiple nodes, where each node has an individual demand shock. When the shocks are multi-dimensional, it is generally not possible to find an *ex post* optimal supply curve. Wilson was none-the-less able to derive first-order conditions for the *ex ante* optimal choices and hence characterise an equilibrium for a general transmission network. The equations are quite complicated, as they depend on the probability distributions of the demand shocks.

Anderson and Philpott (2002b) characterize the residual demand by its market distribution function, which implicitly determines the contour of the residual demand curve for each prob-
ability level. This general formulation allows for equilibria that are \textit{ex ante}, but not \textit{ex post}, optimal and residual demands that are not restricted to the linear shifts of the supply function equilibrium model. Analogously, one can introduce offer distribution functions, which implicitly determine the contour of each producer’s offer curve for each probability level, to analyse mixed strategy equilibria in electricity auctions and other multi-unit auctions (Anderson et al., 2009).

3 Competition policy

Newbery (1998) analytically solves for the supply function equilibrium for two symmetric duopolies with constant marginal costs $c$, linear demand $D(p) = a - bp$, and additive shocks. In the appendix, this solution is generalised to $N > 2$ symmetric firms with market capacity $\overline{Q}$:

\[
\frac{x}{\gamma} = \frac{N (p - c)}{(N - 2) c} \left( (N - 1) \left( \frac{1}{N \gamma} \frac{c}{p - c} \right)^{(N-2)/(N-1)} - 1 \right),
\]

where $x = Q/\overline{Q}$ is the load factor (output as a fraction of total capacity) and $\gamma = cb/\overline{Q}$ is the elasticity of demand at the efficient price $c$ (i.e. the marginal cost) when the demand shock $\varepsilon$ is such that the linear demand at this price equals the market capacity. The relation is based on the upper supply schedule, so it is a worst case scenario. If the price cap is sufficiently low, so that it binds, the market will have a more competitive unique equilibrium. We can use (2) to prove that it is beneficial for competition to connect two symmetric markets. This will double the capacity, demand, including its slope, and the number of producers in the integrated market. Hence the dimension-less constant $\gamma = cb/\overline{Q}$ is the same in the integrated market as for the separate markets, but $N$ has doubled. Thus mark-ups, $\frac{p-c}{c}$, will be reduced for every load factor $x$. But if instead the same firms operate on the two separate markets, market integration neither influences $N$ nor mark-ups as a function of the load factor. Hence, large cross-elasticities for consumption in adjacent delivery periods will not make bidding more competitive if demand shocks in the two periods are perfectly correlated.

As illustrated in figure 6, the mark-up results in under-production, which gives a deadweight loss, DWL. In the appendix we relate the mark-up $\frac{p-c}{c}$ to the relative DWL (i.e. expressed as a fraction of the short-run industry profits) $\omega$, so that (2) can be written as follows:

\[
(2N\omega)^\frac{1}{N-1} = x \left( \frac{N-2}{N-1} \frac{N(1 + 2\omega)}{(N - 1)} - 1 \right)
\]

We note that the relative DWL is independent of the demand elasticity, $\gamma$, for constant marginal costs. In figure 7 the relation in (3) is plotted as a function of the load factor for 5 and 10 firms. Using production costs of the English Pool in 1988/89 (Green and Newbery, 1992), we have also calculated corresponding results for symmetric SFE of a realistic market. When marginal costs
Figure 6: *Mark-ups lead to under-production, which gives a deadweight loss (DWL).*

are increasing, the relative deadweight losses depend on the demand elasticity. The short-run demand elasticity of electricity markets have been estimated to be in the range 0.05 to 0.4. Thus the results for the English Pool have been calculated for a wide interval: \( b = 0.1 \) and \( b = 0.5 \) correspond to \( \gamma = 0.08 \) and \( \gamma = 0.4 \). We see that typically the relative DWL is increasing with the load factor. The reason is that the mark-up is normally convex in output. This convexity is less pronounced for increasing marginal costs (which makes supply functions steeper) and more elastic demand, when the slope of competitors’ supply functions has relatively less influence on mark-ups in comparison to the slope of the exogenous demand, and the relative DWL curve is flatter in that case. In the cases studied here, five symmetric firms are enough to keep relative DWL below 1% if the load factor is below 50%. This corresponds to a Herfindahl-Hirschman index (HHI) of 2000. Ten symmetric firms (HHI = 1000) are needed to keep relative DWL below 1% when the load factor reaches 90%.

In practice it is the average DWL that matters. Considering the load duration curve for the market in England and Wales, Green and Newbery (1992) estimate the average DWL relative to average short-run profits to 1.3% for five symmetric producers and \( \gamma = 0.2 \). With two producers, and otherwise unchanged market conditions, the average DWL relative to average profits is

---

6 von der Fehr et al (2005) estimate the consumer demand elasticity at 0.3 from the high price episode in Norway following hydro shortages, while Patrick and Wolak (1997) cite Borenstein and Bushnell’s (1997) estimates from California of 0.1-0.4, and estimate the short-run demand elasticity by the water industry in England as 0.05-0.27 (depending on the time of day and hence price level). Other individual industries had lower elasticities.

7 The HHI is defined as the sum of the squared percentages of market shares, so 10 firms with 10% each would have an HHI of 10x100=1000. See Tirole (1988) for more details.
Figure 7: Dependence of deadweight loss (DWL) on load factor and number of symmetric firms ($N$).
significantly higher, 8.5% – 9.8% (depending on the slope of the demand curve). Green and Newbery (1992) also calibrate their model to consider different entry scenarios in 1994. All entry is with high-efficient combined cycle gas turbines (CCGT), while the oligopoly is mostly unchanged for the more costly and price-setting coal-fired stations, although incumbents replace some of these stations by CCGT. With five symmetric producers, the market is sufficiently competitive to discourage entry. However, they estimate that there will be significant entry for a duopoly market with $\gamma = 0.2$. Even if the duopoly of the price-setting units is by and large unaffected by the entry, the output of these units is reduced, so that market shares for the two incumbents reduce to 26% each. This also reduces the strategic mark-ups and the average DWL relative to average profits is reduced to 4.9%. Green and Newbery (1992) show that duopoly with entry results in lower concentration and lower average DWL relative to average profits when producers bid below the upper supply schedule.

With asymmetric firms mark-ups often become asymmetric. To model strategic bidding of 153 hydro-producers in Norway, Holmberg (2007) assumes constant marginal costs and asymmetric capacities. The result is that small firms with less capacity and less market-power offer their capacity with lower mark-ups compared to large firms. This implies that the capacity of small firms will bind at lower prices compared to large firms. Hence, many firms in the market compete for the marginal offer at low prices, but the number of competing firms decreases as the price increases. Thus the asymmetric market is more competitive for low demand shocks compared to a symmetric market with a similar market concentration index (HHI index), but less competitive for high demand shocks (see figure ??).

In a symmetric pure-strategy equilibrium of a market with symmetric producers, production is always efficient for a given total output, so that the total production cost of the market is minimized. This is also true for an asymmetric market with identical constant marginal costs. But not if mark-ups are asymmetric and marginal costs are increasing, because then costly production from a low mark-up producer will be accepted instead of less costly production from a high mark-up producer. Figure 8 shows an asymmetric supply function equilibrium for five uncontracted strategic producers in the England and Wales market during 1999. The costs of the strategic producers were approximated by quadratic cost functions as described in Green (1996) and Baldick et al. (2004). The nuclear power producers, BNFL and British Energy, were assumed to be non-strategic and the offers of their must-run plants were assumed to be inframarginal. Subtracting their contribution, the maximum demand shock facing the strategic producers was set to $D(p) = 36 - 0.1p$. The equilibrium, which corresponds to the upper supply schedule, was originally calculated by Anderson and Hu (2008a) and is replicated using an alternative numerical method by Holmberg (2009a). The market shares and market concentration index
depend on the price; HHI varies in the range 2000-5000. To illustrate the production inefficiency we have highlighted the marginal costs of each producer at the price £30/MWh in Fig. 8.

We use the asymmetric SFE to calculate welfare loss relative to profits. The results are presented in Figure 9. We see that the calculated welfare loss relative to profits due to inefficient production is largest around the output 4 GW (4,000 MW) of the strategic producers. At this point National Power and PowerGen have significantly larger output and mark-ups compared to AES, Eastern and Edison Mission Energy that start production at a higher marginal cost. Apart from the peak at 4% the production inefficiency is fairly stable around 2% in the mid-range of the output. It is smaller for low and high outputs. Figure 9 also shows that the dead-weight loss caused by under-production is U-shaped. The curve is very steep for high outputs when the capacity constraints of AES and Edison Mission Energy starts to bind. The high relative dead-weight loss for low outputs can be explained by only National Power and PowerGen competing in the market for low outputs, which gives relatively high mark-ups. Moreover, as competition is increasing for higher outputs, the offer curves of National Power and Power Gen are locally concave for small outputs, see figure 8, which gives the relative welfare loss a decreasing shape.

In the asymmetric market with five strategic firms, all producers have similar production costs and capacities. Even so, the DWL due to inefficient production is of the same order of magnitude as the DWL due to underproduction. Thus one would expect the production inefficiencies to dominate if the asymmetry is larger, especially if the smallest firms have a relatively large fraction of the stations with high production costs.
3.1 Forward contracting

It is well-known that forward sales mitigate the market power of electricity producers. This has been shown both empirically (Wolak, 2000; Bushnell et al., 2008) and theoretically (von der Fehr and Harbord, 1992; Newbery, 1998a; Green, 1999a). Chao and Wilson (2005) show that the mitigation can be even more successful with option contracts. Given the large potential market power that a low demand elasticity and concentrated market structures confer on incumbents in real-time, contracts are potentially very important. As generators in the world’s restructured electricity markets tend to sell a large fraction of their output in the contract market (Green, 1999a; Bushnell, 2007) the effect in reducing deadweight loss is very welcome. But high contract cover is not true for all markets. For example, one of the alleged reasons for the failure of the California Power Exchange was the very limited use of forward contracting by market participants and a high reliance on spot market trading (Bushnell, 2004).

Mark-ups in the real-time market only influence the revenue from sales net of forward contracting. Hence, it is the residual demand net of forward contracts that are relevant for a profit maximizing producer. It is immediate to show (e.g. Anderson and Hu, 2008c; Holmberg, 2008b)
that a producer’s optimal supply curve is given by the natural generalisation of equation (1):

$$\frac{p - C'(q_i(p))}{p} = \frac{-1}{\gamma_{i}^{\text{net-res}}}.$$  

where $q_i$ is the producer’s total output including forward sales and $\gamma_{i}^{\text{net-res}}$ is the elasticity of its residual net-demand. As before, a producer offers positive net-supply with positive mark-ups in the real-time market. If a producer has negative net-supply, i.e. it has to buy back electricity in the real-time market, then it will use its market-power to push down the price. Hence, mark-ups are negative for negative net-supply. Mark-ups are zero at the contracting point where net-supply is zero. These statements are formally proven by Newbery (1995, 1998a) and Anderson & Hu (2008c) and used to identify contract positions by Hortacsu and Puller (2008). Figure 10 shows how the equilibrium in figure 4 is changed by forward contracting.

We note that adjusting contract cover closer to the expected output reduces mark-ups at that point, as mark-ups at the contracting point are zero. Hence, mark-ups would be small if all producers were almost fully contracted. To analyse the effect of the forward contracts, we revisit the calculations in figure 7, but in figure 11 we now assume that half of the expected output is sold in the forward market.  

---

Note that the amount of contracting is increasing with respect to the expected output. Thus a new SFE is calculated for each expected output level. The standard deviation in the demand is assumed to be small for each expected output level so that the forward price approximately equals the spot price for the expected output level.
Figure 11: Dependence of deadweight loss (DWL) on load factor and number of symmetric firms (N).
Comparing 7 and figure 11 we see that the welfare losses are reduced by a factor 4-10 when half the expected output is sold in the forward market. The reduction of the welfare losses is likely to be even larger for most real electricity markets, where the contract cover is more like 90%. Hence, with significant forward contracting, sufficient competition in electricity markets can be maintained with only a few number of firms.

Obviously high contract cover reduces producers’ market-power, so why do they sell in the forward market and how can market regulators stimulate such sales? One way is for regulators to mandate a sufficient volume of forward sales. According to Bushnell (2007) this is an important factor when explaining forward sales in many U.S. markets. This is becoming more common in Europe as well. For example, the European Commission has recently conditioned proposed mergers on that the merged electricity producers have to sell parts of their capacity in virtual power plant auctions. This temporary divestiture is equivalent to the sale of a package of forward contracts (Ofgem, 2009). It can also be in producers’ own interest to sell in the forward market if they are risk-averse.

Forward contracting can also be driven by arbitrage opportunities. The risk in the electricity market is mainly non-systematic, i.e. the correlation between electricity prices and the stock market as a whole is weak. Hence, one would expect that the risk premium for risks carried in the electricity market would be small, because speculators can, at least in theory, completely eliminate non-systematic risks by holding a well-diversified portfolio (Hull, 1997). However, many empirical studies suggest that risk premia can be significant in electricity markets as in Longstaff’s and Wang’s (2004) study of the PJM market (U.S.). Kristiansen (2007) reaches a similar conclusion for Nord Pool, the electricity market of the Nordic countries. Anderson and Hu (2008b) use a supply function equilibrium model to show that such a risk premium can arise when strategic retailers prefer to buy in the forward market in order to increase producer’s forward sales, which lowers their mark-ups in the real-time market. Similarly, such a risk-premium can occur if consumers have to buy in the forward market, see our discussion on imbalance penalties in Section 4.4. Green (1999b) motivates the risk premium by noting that retailer’s have thinner margins and so are more risk-averse than producers. Given that much trading is bilateral and not marked to market as in futures markets, the risks of bankruptcy are very real for retailers (and have certainly occurred in Britain, Texas and California) when there are sudden very large wholesale price movements, so retailer risk aversion is understandable.

But forward contracting in markets with imperfect competition is also driven by strategic mechanisms. Newbery (1998a) shows that producers may use contract sales to keep output high and spot prices low to deter entry. The risk of keeping prices high is that entrants will be attracted by the observation that the incumbents are able and willing to maintain profitability,
but once they enter their capacity will remain to suppress prices for the next 20-40 years. Indeed, high prices encouraged a “dash for gas” and excess entry into the English market in the 1990s, and the incumbents appeared to have tacitly colluded to maintain prices while selling off plant, only for those prices to fall dramatically as the new owners, with smaller market shares, bid more aggressively and caused prices to collapse in 2000-01.

Financial trading is anonymous in most power markets, and a firm’s forward and futures positions are not revealed to competitors. Still, producers may be able to make rough estimates of changes in competitors’ total futures positions by analysing changes in the turn-over in the forward market and the forward price. That competitors can deduce this information gives producers another strategic incentive, which was first illustrated by Allaz and Vila (1993). In their two-stage Cournot model, producers use forward contracts as a first-mover device. By selling to consumers in the forward market, a producer can shift consumer’s real-time demand (net of forward contracts) inwards. If competitors observe or infer what is happening they would respond by shifting their optimal output inwards. This makes it possible for a producer to increase its market share and profit by selling in the forward market. However, all producers have the opportunity to sell in the forward market and they all have incentives to do so. Hence, instead of increased market-shares, strategic contracting just leads to more competitive bidding and lower profits for producers, which is good for welfare but bad for producers. In the limit as firms recontract (assuming they can observe earlier contract positions) the market becomes perfectly competitive. Actual markets have something between 3-8 rounds of contracting (as measured by the churn rate, Ofgem, 2009) but most of these subsequent rounds are between traders, retailers or generators adjusting their positions at the margin and do not amount to additional contracts modelled by Allaz and Vila.

Allaz and Vila’s result appears attractive for the designers of electricity markets. Unfortunately the results are not robust to alternative specifications of competition. Murphy and Smeers (2005) show in a multi-stage Cournot model that if capacity is chosen knowing that in the subsequent periods contracting will make competition more intense and profits lower, firms will invest less in order to keep capacity tight and prices higher in compensation, just as Kreps and Scheinkman (1983) showed that intense price (Bertrand) competition in the post investment period would lead to Cournot choices of capacity that would constrain output and support Cournot equilibria. Mahenc and Salanié (2004) show that strategic forward contracting can have the opposite effect if contracts are observable and producers compete in prices rather than quantities. Their result is that producers have incentives to buy in the forward market (negative contracting), which will push up mark-ups. While squeezes such as this have been observed in electricity markets (and were the subject of a competition inquiry in Britain in the case of Edison
they are unusual (perhaps because they attract the attention of regulators).

Unlike the Bertrand and Cournot equilibrium, the supply function equilibrium is *ex post* optimal to a range of additive demand shocks. A linear shift of the net-demand curve in the real-time market, due to increased forward sales, has no influence on competitors’ optimal offers. To influence competitors’ bidding in this model, the slopes of competitors’ residual net-demand curves have to change. Moreover, they have to be able to predict these changes from observed changes in forward sales. Holmberg (2008b) shows that a strategic producer will use the forward market in order to commit itself to less elastic net-offer curves in the real-time market, because this will increase competitors’ mark-ups, so that their output curves shift inwards, and the strategic producer can increase its market-share on their expense. Whether such a commitment of a strategic producer results from more contracting, which is pro-competitive, or less contracting, which is anti-competitive, depends on the curvature of the marginal cost and residual demand curves (Holmberg, 2008b). When the producer increases its forward sales, the marginal cost curve as a function of the net-supply (net of forward contracts) will shift inwards, and so will the net-supply curve. The inward shift of the net-supply curve is largest at points where the marginal cost curve is steep (as illustrated in figure 12) and at points where the mark-up is small. Hence, if marginal costs are convex and the residual demand is concave, selling in the forward market will shift the offer curve more inwards at high prices compared to low prices. This will make the producer’s net-supply curve less elastic for all prices and competitors’ optimal net-supply curves will become less competitive as well. However, the result is the opposite for concave marginal costs and convex residual demand, provided the producer’s capacity is non-binding (pro-competitive effects cannot be ruled out for all prices if production capacities are binding). If both marginal costs and residual demand are linear (which requires that firms offer linear SFs as in the linear supply function equilibrium model by Green, 1999a) and the market capacity is non-binding, then strategies are neutral; changed forward sales do not directly influence competitors’ bidding. Empirically, marginal costs and offer curves are typically convex. Convex offer curves result in convex residual demand if demand is convex or sufficiently inelastic. Hence, the Allaz and Vila effect is non-robust in electricity markets. Moreover, it is uncertain how strong this effect is as competitors’ forward positions are not directly observable.

At the margin, increased forward sales of a producer can also result in less forward sales for competitors (rather than increased consumer contracting). Green (1999a) shows that this makes competitors’ bidding less competitive. But in equilibrium all producers will increase their forward sales, which makes the market more competitive. The pro-competitive effect increases with liquidity provided by competitors in the forward market. As competitors can always observe

\[9^{9}\] See e.g. http://www.google.co.uk/search?q=edison+mission+market+abuse&sourceid=navclient-ff&ie=UTF-8
Figure 12: The marginal cost curve as a function of net-supply (supply net of forward contracts) shifts inwards when forward sales increase, and there is a corresponding inward shift in the net-supply curve. If marginal costs are strictly convex, the inward shift tends to increase with the price, and this makes the net-supply curve less elastic for all price levels.

their own contracting level, observability is not an issue in this case. Holmberg (2008b) verifies this pro-competitive effect for general cost functions in a duopoly market.

3.2 Empirical results

With many firms in the market, the supply function equilibrium model predicts that electricity prices are near the marginal cost until the capacity constraint binds, where there is a steep increase in the price (see figure 4). This phenomenon is called “hockey-stick bidding”, which has been observed in U.S. markets (Hurlbut et al., 2004). Figure 13 shows the same effect observed on a European power exchange, where the price-cost mark-up only becomes appreciable as the available supply falls to less than 110% of the estimated demand (so that the reserve margin falls below 10%).

The supply function equilibrium model also has empirical support in that two empirical studies of ERCOT (a balancing market in Texas) suggest that the bids of the two to three largest firms do indeed match the first-order condition of the continuous SFE (Hortacsu and Puller, 2008; Sioshansi and Oren, 2007). Niu et al. (2005) are also able to replicate observed market prices in this market using a model with linear supply function equilibria and fixed forward contracting. Willems et al. (2009) are able to replicate mark-ups in the German electricity market by using a SFE model with calibrated contracting cover. They also find a similar fit for a Cournot model.

\[\text{Price, marginal cost} \]

\[\text{Inward shift of } S(p) \]

\[\text{Inward shift of } MC\text{-curve} \]

\[\text{Net-supply (net of forward contracts)} \]

---

\[10\] The EU Sector Inquiry gives many excellent examples of similar scatter diagrams, see London Economics (2007).
Figure 13: Price-cost mark-up observed on a European power exchange.

if they allow for a higher contracting level.

There are other empirical approaches that test the underlying assumptions of the SFE model. Wolak (2003a) backs out the unobserved underlying cost and contract positions of generators bidding into the Australian market. He cannot reject the hypothesis that producers use pure and static strategies to maximize their expected profit with respect to an uncertain smoothed residual demand. Hence, his results indirectly support the supply function equilibrium model. Sweeting (2007) makes the assumption that producers’ contract cover is 80% (or larger) and is able to characterize the various phases of market evolution and the exercise of market power in the English Electricity Pool. He concludes that the behaviour of the two largest generators was consistent with either tacit collusion or an attempt to raise the Pool prices, so that they could negotiate higher forward prices and sale prices of the plants they were divesting.

Wolak (2003b) uses observed individual bid and offer curves to estimate the elasticity of the residual demand for 5 large producers in the California market. From these results he can estimate potential mark-ups using (1). On average the potential mark-ups were 15% in 2000, which was 3-5 times higher than in 1998 and 1999.
4 Regulation of electricity auctions

4.1 Should offer curves be disclosed?

Some markets, e.g. in Britain, New Zealand and Australia, disclose individual offer curves with some time delay. In this case, it is easy to calculate the residual demand elasticity of each producer at the clearing point as in Wolak (2003b). This greatly simplifies market surveillance for market regulators and competition authorities, as no information about production costs is needed to calculate potential mark-ups. The advantages are especially large for markets where production costs or opportunity costs are not transparent, which is the case in hydro-dominated markets where opportunity costs are determined by the producers’ prognoses of future inflows to the reservoirs and future electricity prices. A disadvantage with disclosing offer curves is that producers can use the disclosed individual offer curves to monitor competitors’ signals and how well they follow an implicit or explicit collusive strategy. But this risk is reduced if only parts of the offer curves around the clearing point are disclosed, which provides sufficient information for the approach by Wolak (2003b), and if disclosure is significantly delayed, perhaps up to one year. Some markets (e.g. APX in figure 1) disclose aggregated supply and demand curves. This makes it possible for producers, who know their own offer curves, to exactly calculate their residual demand, so that they can monitor their competitors’ aggregate offers. But this type of disclosure does not give regulators the same possibility for market surveillance. Unless the disclosure of aggregated offers/bids are delayed by a long time, it seems that it facilitates tacit collusion without improving regulators’ monitoring possibilities.

4.2 The price cap and investment

Normally the price cap is set to the value of lost load (VOLL). Under perfect competition this will theoretically lead to optimal investment (Stoft, 2002), as long as investors are risk-neutral. Even if markets are competitive, peaking power plants can still cover their fixed costs during the rare events when there is an electricity shortage, so that the market is cleared at the price cap. An advantage with this regulation is that producers always want to offer all of their available production capacity to the market, because it is always better to offer capacity at the price cap than to withhold it, at least in a static equilibrium (Holmberg, 2008a). To perfectly hedge against uncertainties in their residual demand, producers can issue call options (or equivalently one-way contracts for differences) for each production unit with a quantity and strike price corresponding to the unit’s capacity and marginal cost (Chao and Wilson, 2005). As suggested by Oren (2005),

11 In the Pool until it was abolished by the New Electricity Trading Arrangements in 2001, and thereafter in the Balancing Mechanism.
a market for call options is a natural way of implicitly introducing capacity payments (discussed below).

Some markets have replaced the high price cap with explicit capacity payments and a lower price cap, \( p_{\text{cap}} \). To get optimal investments in a competitive market, the capacity payment should equal LOLP* \( (VOLL - p_{\text{cap}}) \), which ensures that producers in competitive markets are paid the same amount on average as in markets without capacity payments and the price cap equal to VOLL. Capacity payments are determined on the basis of LOLP for every period, and although they will normally be very low, for some hours they can be very high when reserve margins are below a critical level. Capacity payments are independent of the bids (but increase exponentially as the reserve margin falls), so the supply function equilibrium can still be used to calculate equilibria in such markets. Hence, an advantage with capacity payments is that producers will bid more competitively if the price cap is sufficiently low so that it binds for the highest bids (see figure 14), and the capacity payment will not compensate strategic producers for the reduction in their mark-ups. This result in lower average prices and reduces over-investment due to excessive entry. However, a problem is that it might be profitable for producers to withhold production from the auction in order to increase LOLP and the capacity payments. This was a problem in England and Wales (Newbery, 1995) and led to a change in the Pool rules so that plant would not influence the calculated LOLP until 8 days after withdrawal. Similar to the issue of call options, an advantage with capacity payments is that it stabilizes risk-averse producers incomes. A major problem with capacity payments is that even with detailed information, it is likely that there will be large errors in the calculated capacity payments, because it is very difficult for a

Figure 14: Reducing the price cap pushes down the equilibrium price for every demand outcome.
system operator to correctly estimate LOLP (Newbery, 1998b).

A potential problem for investments is that producers may not trust market regulators to keep a high price cap and high capacity payments during a period of extended electricity shortages, as it would lead to very high electricity prices. Although home consumers can contract to reduce the adverse consequences of the resulting price spikes, there might be political pressure on regulators to lower the price cap or capacity payments, in order to push down electricity prices as illustrated in figure 14, even if such an irrational measure would exacerbate the shortage, especially in interconnected markets, where neighbouring markets with electricity shortage need to compete for reserve capacity (Stoft, 2002).\textsuperscript{12} Hence, even if the price cap is set to VOLL, producers might under-invest due to the regulatory risk that the price cap might be lowered in the future. As with granting independence to central banks, it should be possible to reduce this potential time-inconsistency problem by making regulators of electricity markets more independent.

Entry and investments will be stimulated by high mark-ups. Mark-ups for production late in the merit-order are to a large extent determined by the price cap and LOLP. Hence, for a given price cap and LOLP, marginal investments in peaking power plants should be less dependent on the number of firms in the market compared to investments in off-peak production. Hence, compared to a competitive market with the same market capacity, one would expect more entry with off-peak power plants into a non-competitive market, which may be true for incumbents' investments as well.

4.3 Restrictions on the offer curves

In the pre-2001 pool of England and Wales, each producer had to stick to the same offer curve during the whole day. In the Australian market producers are restricted to choose ten price levels per production unit, which they must maintain during the whole day, although the quantities offered at the chosen prices are allowed to differ in each delivery period. Such auction bidding rules can constrain the set of punishment and signalling strategies available to strategic producers, which may discourage collusion. For example, a larger demand variation for a given offer curve widens the range of bids that are price-setting with a positive probability. This reduces the risk that parts of the offer curves that are only price setting for out-of-equilibrium events are used as costless signals to coordinate collusion or costless threats to discourage deviations from a collusive equilibrium (Klemperer, 2004). Figure 15 shows how the lower part of an offer curve, where offers are always inframarginal, can be chosen to discourage competitors from undercutting

\textsuperscript{12} The highest price spike in the English Pool occurred when the French market needed to import power instead of its normal export behaviour, and at the same time a British nuclear power station tripped.
the lowest realized equilibrium price, because in this case the output of the producer will remain fixed for such an out-of-equilibrium event. It is therefore clear that restrictions on the offer curves can make bidding more competitive in one-shot games. But welfare is reduced if offers are too constrained, as this may lead to inefficient production if a unit’s costs vary during the day.

Auction rules can also constrain the shape of the offer curves. The supply function equilibrium model assumes that offers consist of smooth supply curves. However, in practice electricity auctions require offers to be stepped or piece-wise linear curves. For administrative reasons there are also quantity multiples, price tick sizes, and restrictions in the number of steps per bidder (see Table 1). von der Fehr and Harbord (1993) and Kremer and Nyborg (2004b) show that a large quantity multiple and small price tick-sizes will encourage bidders to undercut each other, as in a Bertrand game, which will result in a competitive equilibrium with zero mark-ups for non-pivotal producers with constant marginal costs. But this mechanism does not work for demand outcomes where producers are pivotal\(^\text{13}\) because he will find it profitable to deviate from such a competitive outcome. Von der Fehr and Harbord (1993) show that the combination of large quantity multiples and small price tick-sizes kills any pure-strategy equilibrium if the demand variation is sufficiently large and at least one producer is pivotal at the highest demand outcome. Instead, as with the Bertrand-Edgeworth model (Edgeworth, 1925), there will be mixed-strategy equilibria with randomized bidding.

Analogously, Anderson and Xu (2004) show that strategic producers in the Australian mar-

\(^\text{13}\) A producer is pivotal if competitors do not have enough capacity to meet market demand in his absence.
Table 1: Constraints on the supply functions in various electricity markets.

<table>
<thead>
<tr>
<th>Market</th>
<th>Max steps</th>
<th>Price range</th>
<th>Price tick size</th>
<th>Quantity multiple</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nord Pool spot</td>
<td>64 per bidder</td>
<td>0-5,000 NOK/MWh</td>
<td>0.1 NOK/MWh</td>
<td>0.1 MWh</td>
</tr>
<tr>
<td>ERCOT balancing</td>
<td>40 per bidder</td>
<td>-$1,000/MWh to $1,000/MWh</td>
<td>$0.01/MWh</td>
<td>0.01 MWh</td>
</tr>
<tr>
<td>PJM</td>
<td>10 per genset</td>
<td>0-$1,000/MWh</td>
<td>$0.01/MWh</td>
<td>0.01 MWh</td>
</tr>
<tr>
<td>UK (NETA)</td>
<td>5 per genset</td>
<td>-$9,999/MWh to $9,999/MWh</td>
<td>£0.01/MWh</td>
<td>0.001 MWh</td>
</tr>
<tr>
<td>Spain Intra-day market</td>
<td>5 per genset</td>
<td>Yearly cap on revenues</td>
<td>£0.01/MWh</td>
<td>0.1 MWh</td>
</tr>
</tbody>
</table>

Market design should randomise their choice of stated price levels. It is unclear whether the large quantity multiples assumed in the von der Fehr and Harbord (1993) model might result in lower average mark-ups for mixed-strategy equilibria, but it is clear that the market participants would bear the cost of uncertainty caused by the inherent price instability. Moreover, randomized bidding will cause production inefficiencies. For example, symmetric producers with strictly increasing marginal costs do not cause any production inefficiencies for a given market output in a symmetric pure-strategy equilibrium, but they do in a symmetric mixed-strategy equilibrium. On the other hand an advantage with mixing is that it widens the range of bids that are price-setting with a positive probability. This decreases the risk that out-of-equilibrium bids are used as costless signals or costless threats. Newbery (1998a) conjectures that the mixing may not be too severe if each producer is allowed to choose many steps in their offer curves. Parisio and Bosco (2003) show that pure-strategy equilibria can occur also in markets with large quantity multiples relative to the price tick size if there is sufficient uncertainty in competitors’ cost functions.

Holmberg et al. (2008) show that if the market design instead has small quantity multiples relative to the price tick size and if the market has lax or no restrictions on the number of steps, then undercutting down to the marginal cost is avoided and the market has a pure-strategy equilibrium that converges to the smooth supply function equilibrium as the number of price levels increases, also for pivotal producers. Hence, if market designers want to avoid inherent price instabilities, they can choose large price tick sizes, small quantity multiples, and lax restrictions on the number of steps per bidder. Under beneficial circumstances, Anderson and
Hu (2008a) show that pure-strategy equilibria exist also for piece-wise linear supply functions, and that they converge to smooth supply function equilibria. Electricity auctions with piece-wise linear offer curves are used in Nord Pool (Nordic countries) and PowerNext (France). This seems to be the easiest way to avoid a market design with the inherent price instability caused by mixed strategies.

Rationing of excess supply at the clearing price is often necessary in multi-unit auctions and so market designs must specify how rationing will take place. This is normally by pro-rata on-the-margin rationing (Kremer and Nyborg, 2004a) in which only the incremental supply at the clearing price is rationed and the accepted share of each producer’s incremental supply at this price is proportional to the size of this increment offered. Holmberg et al. (2008) note that this rationing rule has the advantage that profits in a one-shot game are maximised when supra-marginal offers (offers that are never accepted in equilibrium) are offered with a perfectly elastic segment along the highest realised equilibrium price. This strategy maximizes the size of a producer’s incremental supply and accepted output at the highest shock outcome. Hence, pro-rata on-the-margin rationing will make it costly to use supra-marginal offers for signalling and threats in multi-unit auctions. This form of supra-marginal offers is also beneficial in supporting the most competitive SFE consistent with producers’ capacity constraints.

4.4 Imbalance penalties in real-time markets

Normally, only flexible producers and flexible consumers, who can quickly change their production/consumption, are allowed to bid in the real-time market. Hence, imbalance penalties give inflexible market participants (who cannot control their consumption/production in real-time) an incentive to make accurate forecasts of their consumption/production and to contract this amount ahead of time. This reduces the imbalances that the real-time/balancing market need to handle. This is an advantage as the balancing market is less liquid compared to the day-ahead market as most production and consumption is inflexible at short notice. Penalty charges can also deter attempts to corner the real-time market. A sufficiently liquid real-time market is crucial if one wants to keep the loss of load probability low, and is also beneficial for competition. But the designs of real-time markets are sometimes flawed in that they punish deviations from contracted sales too much. This gives producers incentives to hoard their balancing power internally, so that they can counteract any internal imbalances on their own to avoid any penalty charges. This phenomenon was apparent in the original NETA design of Great Britain (Roques et al., 2005). An interesting question is how penalty charges can be combined with a sufficient liquidity in the real-time market. If penalty charges were capped at the additional cost that would apply if the producer’s imbalance were counteracted by its own offers instead of cheaper accepted offers from
Figure 16: Producer surplus in uniform-price and pay-as-bid auctions.

competitors, it would always be profitable to offer all flexible capacity to the real-time market.

Consumers are typically inflexible, so imbalance penalties will give them incentives to contract ahead of time. Market prices are approximately convex with respect to the system imbalance, so consumers tend to contract slightly more than they need on average. As aggregate imbalances are typically small they should result in small mark-ups in the real-time market. However, it is uncertain to what extent they would spill-over to small mark-ups in day-ahead and forward markets, where most power is traded. The problem is that imbalance penalties discourage arbitrage for inflexible market participants and speculators. Hence, mainly flexible producers can utilize arbitrage opportunities between the day-ahead and real-time market, but strategic producers may not want to utilize them fully. Due to the risk-premium, individual consumers may prefer to buy in real-time and that all other consumers buy in the forward market to reduce real-time mark-ups. But as a group consumers should gain if they all have to buy in the forward market.

4.5 Pay-as-bid auctions

The pay-as-bid auction is used in most treasury auctions around the world, but it has so far been less popular in electricity markets. One exception is England & Wales, which switched to such a format in 2001 for the balancing mechanism (the only remaining auction “market” as the day-ahead market is effectively an OTC market). Italy seems to become another exception, a recent law dated January 2009 says that their day-ahead market should switch to a pay-as-bid format.

If producers would bid with the same offer curves in pay-as-bid auctions as they use in uniform-price auctions, then average prices in electricity markets would be significantly lower with the pay-as-bid format as illustrated in figure 16. However, as pointed out by Kahn et al. (2001),
Wolfram (1999) and others, this naïve assumption is not reasonable, as strategic producers will change their optimal bidding strategies when the auction format changes. Moreover, theoretical and empirical results applicable to treasury auctions show that rankings of the auctioneer’s revenue are ambiguous for the two formats (Ausubel and Cramton, 2002). In this auction buyers have common/affiliated uncertain values given by the value of the security in the secondary market. Hence, these results should be applicable to day-ahead auctions in electricity markets, where producers’ opportunity costs of selling in a later market are common/affiliated.

Electricity prices went down in Britain after their real-time market switched to a pay-as-bid format. But Evans and Green (2005) use a supply function equilibrium model to control for investments and market concentration in Britain and they conclude that low electricity prices after the introduction of pay-as-bid pricing in 2001 can be explained by added capacity as well as forced and voluntary divestitures rather than the changed auction format. An experiment by Rassenti et al. (2003) even suggests that average prices might be higher in pay-as-bid auctions, at least if demand is certain.

Game-theoretic models of real-time markets with static strategies, however, have so far concluded that short-run average prices are lower with the pay-as-bid format, at least if contracting is neglected. Fabra et al. (2006) came to this conclusion using an auction model, where producers choose one price for all of their capacity, as did Son et al. (2004) with a similar model. Holmberg (2009b) reaches similar conclusions for a supply function equilibrium model. His results suggest that switching from a uniform-price to a pay-as-bid format will be most beneficial for the auctioneer/consumers when the risk of power shortage is neither extremely small nor extremely large and when competition is imperfect without being extremely bad. Under perfect competition producers’ expected profits are the same regardless of the auction format. Wang and Zender (2002) use a model of strategic bidding, which is essentially a supply function equilibrium model, to compare auction formats in treasury auctions. They also conclude that the auctioneer would prefer the pay-as-bid format.

These theoretical results were all derived for oligopoly markets with inelastic demand and a fixed price cap. Elastic demand in pay-as-bid auctions creates some difficulties. As with Bertrand models (Maskin, 1986), the question is how much of the excess demand at lower offered prices will remain at higher prices. Federico and Rahman (2003) assume that demand depends on the highest accepted offer in the market though much of the demand is met at lower prices, because of the pay-as-bid mechanism. The corresponding assumption in Bertrand models is called parallel rationing (Maskin, 1986), which is equivalent to the assertion that demand does

14Indeed, in single object auctions the Revenue Equivalence theorem implies that payments to producers will be the same under either format. See e.g. Klemperer (2004).
not depend on income (Levitan and Shubik, 1972). The result of Federico and Rahman’s (2003) assumption is that the non-strategic demand side of their model is the same in the pay-as-bid and uniform-price format, which simplifies their analysis. Using a supply function equilibrium model, Federico and Rahman (2003) show that consumer surplus is larger in pay-as-bid auctions compared to uniform-price auctions for both monopolistic and perfectly competitive producers. However, output and total welfare are sometimes lower in the pay-as-bid format and these models do not discuss long-run equilibrium in which capacity is endogenous.

Under suitable conditions there will be a pure strategy NE in pay-as-bid auctions, where each producer chooses a smooth optimal supply curve given the properties of its residual demand. But compared with uniform-price auctions, it is normally harder to derive a supply function equilibrium for pay-as-bid auctions. *Ex post* it is always optimal to offer all accepted bids at the same price, the marginal bid. Hence, *ex post* optimality for a range of demand shocks in a pay-as-bid auction would require horizontal offers. As with the Bertrand NE, pure-strategy equilibria with horizontal offers are possible if capacity constraints are non-binding and marginal costs are constant (Wang and Zender, 2002). But this is an exception, typically equilibrium offers to the pay-as-bid auction are not *ex post* optimal. Hence, as with the generalisations of the SFE model discussed in Section 2.1, optimal offers will normally depend on the probability distribution of the demand shock. The pay-as-bid format also implies that the offer price of one unit does not influence the profit from the producer’s other units, as long as the units are offered in merit-order. Hence, the expected profit from each unit can be optimized independently. Rather than choosing the offer curve such that the profit is maximized for each shock outcome, which is the case in the uniform-price format, producers in the pay-as-bid format choose an offer curve so that the expected profit from each production unit is maximized (given the properties of the residual demand).

Let \( F(\varepsilon) \) be the probability distribution of the demand shock and \( f(\varepsilon) \) the density function of this shock. Assume that a unit \( q^* \) of the producer is offered at the price \( p^* \) and that this offer is marginal for the shock outcome \( \varepsilon^* \). Then the optimal offer of a production unit is given by the following relation (Holmberg, 2009b, Anderson et al., 2009):

\[
[p - C'(q^*)] f(\varepsilon^*) \Delta \varepsilon - [1 - F(\varepsilon^*)] \frac{\Delta \varepsilon}{\frac{\partial D_{\text{res}}(p^*)}{\partial p}} = 0. \tag{4}
\]

The first term is the expected marginal gain from the unit \( q^* \) if its offer is lowered, so that the range of demand shocks for which the offer is accepted increases from \( \varepsilon^* \) to \( \varepsilon^* + \Delta \varepsilon \). Hence, the probability that the bid of unit \( q^* \) is accepted increases by \( f(\varepsilon^*) \Delta \varepsilon \), which considering the unit’s mark-up, gives the expected marginal gain \( (p - C'(q^*)) f(\varepsilon^*) \Delta \varepsilon \). The second term is the
loss associated with increased sales. To achieve this the offer price is decreased by the amount
\[ \Delta p = \frac{\Delta \varepsilon}{|D_{res}(p^*)|} \]
which results in the marginal loss \[ [1 - F(\varepsilon^*)] \frac{\Delta \varepsilon}{|D_{res}(p^*)|} \] where \( 1 - F(\varepsilon^*) \) is the probability that the offer of unit \( q^* \) is accepted. The optimal offer price \( p^* \) is determined by the requirement that the marginal gain and marginal loss exactly balance each other. Moreover, for a profit maximum, we have the necessary second-order condition that the marginal loss must increase faster with respect to \( \varepsilon^* \) compared to the marginal gain. In contrast to the second-order condition for uniform-price auctions, this is quite a restrictive condition. It is straightforward to show by means of (4) that this will require the hazard rate of the demand shock \( \frac{f(\varepsilon)}{1 - F(\varepsilon)} \) to be decreasing with respect to \( \varepsilon \) at points where the marginal cost is locally constant (Holmberg, 2009b, Anderson et al., 2009). This restriction on the probability distribution rules out many demand shock distributions that one would encounter in practice, including the normal distribution. An exception, where pure-strategy SFE exists also for locally constant marginal costs is the Pareto distribution of the second-kind, which Holmberg (2009b) uses in his calculations.

In uniform-price auctions it is optimal to offer the first unit at marginal cost, also for strategic producers (see figure 4). In pay-as-bid auctions such bidding would result in zero profits from the first unit and they are therefore generally offered with a mark-up. Hence, for low demand outcomes prices are generally higher in pay-as-bid auctions (see right graph of figure 17). If demand is sufficiently inelastic so that the highest bid is determined by the price cap, then the highest bid in both auction formats will be equal to the price cap. In uniform-price auctions the highest bid is price-setting for the whole market, so all producers get this price. In pay-as-bid auctions on the other hand, the average of the price paid to accepted offers (the system buy price) is generally lower than the highest bid. Hence, for high demand outcomes, the system buy price is lower than in uniform-price auctions. In summary, the price variation (the difference between the lowest and highest system buy price) is lower in pay-as-bid auctions when there is a pure-strategy equilibrium. This has also been confirmed in the experiments by Rassenti et al. (2003).

There is, however, a considerable risk that the second-order condition is not satisfied in balancing markets. Garcia and Kirschen (2006) find that system imbalances in Britain are approximately normally distributed. Given the central limit theorem, this is not surprising as the system imbalance results from a large number of actions by different market participants acting independently. Second, marginal electricity production costs are roughly stepped, i.e. approximately locally constant. Hence, one can expect that offer curves submitted to electricity auctions with pay-as-bid formats are at least partly randomized. Genc (2009) shows that mixtures over horizontal offers occur in discriminatory divisible-good auctions for cases with constant marginal costs and uniformly distributed demand (see left graph of figure 17). His equilibrium is
unchanged if producers are restricted to choose one price for their whole capacity as in the auction model by Fabra et al. (2006). These strategic mixtures are essentially Bertrand-Edgeworth equilibria (Edgeworth, 1925), but with uncertain demand. Fabra et al.’s (2006) comparison with mixed-strategy equilibria in uniform-price auctions indicates that the auctioneer prefers the pay-as-bid format also when producers use mixed strategies. Anderson et al. (2009) generalize the results by Genc (2009) by considering mixtures for increasing marginal costs. In this case, mixed strategy equilibria will cause production inefficiencies. With increasing marginal costs and a sufficiently low price cap there will be new types of mixed strategy equilibria where some of the offer curves have strictly increasing parts (see middle graph of figure 17).

A general assumption of the theoretical analyses is that firms are risk-neutral. Introducing risk aversion does not change the SFE of a uniform-price auction, as firms receive the best price for every demand outcome, given the bids of competitors. A risk-averse firm in a pay-as-bid auction, however, would put less weight on high-demand outcomes when profits are high and more weight on low-demand outcomes when profits are low. Hence, given the offers of competitors, risk-averse firms decrease their offer prices to increase their sales and profits for low-demand outcomes. Intuitively this would also be true in equilibrium. It appears that with risk-averse bidders, the advantages (from the auctioneer’s viewpoint) of pay-as-bid auctions are likely to increase.

As a final remark on the game-theoretic analysis we must caution that possible lower short-
run prices in the pay-as-bid auction may not be true in a long-run equilibrium in which investors need to earn a normal return on their investment; they may under-invest in market designs with discriminatory pricing, leading to higher long-run average prices.

An advantage with pay-as-bid auctions compared to uniform-price auctions is that all inframarginal bids are price-setting. Hence, the risk that out of equilibrium bids are used as costless signals or costless threats is lower in pay-as-bid auctions compared to uniform-price auctions (Klemperer, 2004). Fabra (2003) and Klemperer (2004) show that the risk for tacit collusion is lower in pay-as-bid auctions compared to uniform-price auctions.

The pay-as-bid format is disadvantageous for must-run plants that have to bid sufficiently low that their bids are surely accepted. But this is less of a problem if they can sell most of their production in forward markets or if the System Operator invites longer-term tenders for such plant. To make the design of real-time markets more favourable to must-run plants one can also allow for non-competitive bids (Klemperer, 2004). These bids are always accepted and they are paid the system buy price. Non-competitive bids are often allowed in treasury auctions. Similar solutions seem sensible for small electricity producers, because producers need good information of the properties of the residual demand to bid intelligently (Klemperer, 2004). Otherwise pay-as-bid auctions might discourage potential bidders who have only small amounts to trade and for whom the cost of obtaining market information might not be worth paying. For the same reason, a major concern is that the pay-as-bid format might deter potential entrants.

One issue with pay-as-bid auctions is that they have multiple prices, so there is no obvious strike price that can be used to clear financial contracts. In a single-node market it should be possible to use a volume-weighted average of accepted bids as a strike price, but after Britain switched to a pay-as-bid format in the real-time market there is no official strike price. This means that most contracts are physical, which have reduced liquidity in forward markets. According to Ofgem (2009) liquidity in British forward markets is 3 times less than in Germany and the Nordic countries. Another concern with the multiple prices associated with pay-as-bid auctions is that they may cause problems for the system operator when calculating optimal dispatches in a transmission-constrained system with locational pricing. In such a system it will also become less obvious how a strike price should be calculated for each node or price zone. Even if the real-time market has a pay-as-bid format, it is still possible to trade forward contracts in a uniform-price auction. Hence, as in many other countries the introduction of a uniform-price day-ahead market could provide strike prices for financial contracts in Britain. Day-ahead markets are normally much more liquid than real-time markets, which also makes them more suitable for providing strike prices.
5 Conclusions

The supply function model describes optimal bidding in wholesale electricity auctions where demand varies and/or is uncertain. It is a static model, in that producers choose their offers curves to maximize profits at each moment (hour or half-hour) given their residual demand at that moment, without regard to bidding in future periods. Empirical studies of bidding in electricity markets by and large support this model and its underlying assumptions, i.e. ex-post (after uncertainties have been realised) producers maximize their profits for each period given their smoothed residual demand curve. But there are exceptions. For example, Sweeting (2007) shows that producers coordinated to offer curves higher than the SFE model would have predicted during a period when producers were divesting plants, in effect tacitly colluding to keep prices higher and thus maintain the appearance of a more profitable market, leading to higher sales values for their divested plant. After divestiture, margins rapidly collapsed (Newbery, 2005).

This paper provides new results that highlight how short-run welfare losses depend on the number of firms in the market and their asymmetry. We show that market integration reduces mark-ups when demand shocks in the integrated markets are not perfectly correlated or when different producers operate on the integrated markets. Mark-ups depend very much on the elasticity of the demand, but we show that social welfare relative to profits is less sensitive to this elasticity. In this case it is mainly the load factor and the number of producers that matters. Without forward contracts, 5-10 symmetric producers are needed to keep the average welfare loss relative to the average industry profit below 1%. This corresponds to a market concentration index (HHI) of 1000-2000. With asymmetric firms, lower market concentration indices are needed to keep relative average welfare losses below 1% due to the production inefficiencies caused by asymmetric mark-ups. These inefficiencies increase with the asymmetry in the market. More research is needed to analyse long-run welfare losses resulting from inefficient investments.

Forward contracting reduces strategic producers’ mark-up incentives. If producers sell half their expected output in the forward market, relative welfare losses decrease by a factor 4-10. This is of the same order of magnitude as if the number of producers would have been doubled. The question is what incentives producers have to sell in the forward market when this decreases their market-power. Sometimes forward sales are forced upon large producers. For example, competition authorities can order them to sell parts of their capacity in virtual power plant auctions. Introducing penalty charges for deviations from contracted power give consumers incentives to buy their expected demand in the day-ahead and other forward markets. Producers’ forward sales are also driven by their risk aversion and by arbitrage opportunities. As retailers have very thin margins they tend to be more risk-averse, which pushes up forward prices, encouraging producers to offer forward sales.
Forward sales can also be driven by strategic contracting, i.e. the producer wants to increase its market-share at competitors’ expense. This mechanism has pro-competitive effects when competitors are marginal buyers in the forward market. Results are less robust when consumers are marginal buyers in the forward market. A market designer can keep this latter effect weak by introducing penalty charges for deviations from contracted power and by not disclosing producers’ forward positions to the market.

Requiring individual offer curves to be disclosed to regulators makes it straightforward to monitor potential mark-ups and market manipulation. This is a great advantage especially for hydro-dominated markets, which are very complicated to monitor, as production decisions are determined by a prognosis of future inflows to the reservoirs and future electricity prices. Public disclosure should be delayed, perhaps by a year, so that it does not facilitate collusion. To reduce signalling opportunities, it may also be beneficial to only disclose parts of the offer curves around the clearing point, which is enough to monitor potential mark-ups.

Combining capacity payments with a low price cap reduces the risk for investors and makes bidding more competitive. However, one has to make sure that this does not provide incentives to withhold power from the auction, and it is difficult to calculate the loss of load probability accurately, which capacity payments are based on.

Restrictions on the offer curves and long-lived bids decrease the possibilities for tacit collusion, but such restrictions may cause production inefficiencies. Large quantity multiples are useful to ensure competitive bidding when no producer is pivotal, but with pivotal producers, such a design will cause inherent price instabilities due to randomized bidding. These cause production inefficiencies and unnecessary uncertainty for market participants.

For given production capacities theoretical results indicate that average electricity prices are lower if real-time markets use the pay-as-bid format instead of the uniform-price format. However, empirically the effect is not significant and for results applicable to day-ahead markets both empirical and theoretical comparisons of the two formats are ambiguous. One advantage with the pay-as-bid auction is that the risk for collusion is smaller, at least in theory. On the other hand, there is a considerable risk that pay-as-bid clearing results in inherent price instabilities. Moreover, detailed market prognoses are required to bid in pay-as-bid auctions, which disadvantages small producers, deters entry and in any case may have no beneficial long-run effect once investment incentives are considered.

References


Anderson, E. J., P. Holmberg, A. B. Philpott (2009), Mixed strategies in discriminatory divisible-good auctions, Mimeo, University of Sydney.


Green, R.J. (1999b), Appendix to ‘The Electricity Contract Market in England and Wales’. Available at editorial website of *Journal of Industrial Economics*.


Press.


Newbery, D.M. (2008a) Predicting market power in wholesale electricity markets, EPRG


Appendix

Consider $N$ symmetric producers with constant marginal costs $c$ and linear demand $D(p) = a - bp$, so that $D' = -b$. The relation in (1) can be written as follows:

$$q_i - (N - 1) q_i' (p - c) = b (p - c).$$

Now, both sides are multiplied by the integrating factor $\frac{-(p-c)^{-1/(N-1)}}{(N-1)(p-c)}$.

$$\frac{-q_i (p - c)^{-1/(N-1)}}{(N - 1)(p - c)} + q_i' (p - c)^{-1/(N-1)} = \frac{-b (p - c)^{-1/(N-1)}}{(N - 1)}.$$

From the product rule we now observe that

$$\frac{d}{dp} \left[ q_i (p - c)^{-1/(N-1)} \right] = \frac{-b (p - c)^{-1/(N-1)}}{(N - 1)}.$$

We can now integrate both sides.

$$q_i (p - c)^{-1/(N-1)} = A - \int \frac{b (p - c)^{-1/(N-1)}}{(N - 1)} dp = A - \frac{b}{(N - 2)} (p - c)^{(N-2)/(N-1)}.$$

Thus

$$q_i = A (p - c)^{1/(N-1)} - \frac{b}{(N - 2)} (p - c) \quad (5)$$

and

$$q_i' = \frac{A (p - c)^{(2-N)/(N-1)}}{N - 1} - \frac{b}{(N - 2)}.$$

Let $\bar{p}$ be the price where the capacity constraint starts to bind. To ensure smooth residual demand for all producers, offers must be vertical at this point (see figure, i.e. $q_i' (\bar{p}) = 0$. Hence,

$$q_i' (\bar{p}) = \frac{A (\bar{p} - c)^{(2-N)/(N-1)}}{N - 1} - \frac{b}{(N - 2)} = 0,$$

which implies that

$$A = \frac{(N - 1) b (\bar{p} - c)^{(N-2)/(N-1)}}{(N - 2)}.$$

Accordingly, (5) can be written as follows:

$$q_i = \frac{b (p - c)}{(N - 2)} \left( (N - 1) \left( \frac{\bar{p} - c}{p - c} \right)^{(N-2)/(N-1)} - 1 \right).$$
Let $\bar{q}$ be the symmetric capacity constraint. Then we get the following relation at $\bar{p}$

$$\bar{q} = b (\bar{p} - c).$$

Hence, (5) can also be written as follows:

$$q_i = \frac{b(p - c)}{(N - 2)c} \left( (N - 1) \left( \frac{\bar{q}}{p - c} b \right)^{(N-2)/(N-1)} - 1 \right)$$

or equivalently

$$\frac{Nq_i}{cb} = \frac{N(p - c)}{(N - 2)c} \left( (N - 1) \left( \frac{1}{N} \frac{c}{p - c} \frac{N\bar{q}}{bc} \right)^{(N-2)/(N-1)} - 1 \right).$$

We now introduce the two dimension-less numbers $x = q_i/\bar{q}$ and $\gamma = \frac{cb}{N\bar{q}}$. Thus the above expression can be simplified to the following:

$$\frac{x}{\gamma} = \frac{N(p - c)}{(N - 2)c} \left( (N - 1) \left( \frac{1}{N} \frac{c}{p - c} \frac{N\bar{q}}{bc} \right)^{(N-2)/(N-1)} - 1 \right).$$  \hspace{1cm} (6)

The next step is to derive a relation between the mark-up and the deadweight loss. In the case when demand is linear, this loss, which is illustrated in figure 6, can be calculated from

$$W = \frac{(p - c)^2 b}{2}.$$  

The total (short-run) industry profit is $\Pi = (p - c)Nq_i$, so the ratio of deadweight loss to the industry profit $\omega$ is

$$\frac{W}{\Pi} = \omega = \frac{b(p - c)^2}{2(p - c)Nq_i} = \frac{\gamma(p - c)}{2xc} \Rightarrow \frac{p - c}{c} = \frac{2x\omega}{\gamma}. \hspace{1cm} (7)$$

Then equations simplify as (6) becomes

$$\frac{x}{\gamma} = \frac{2Nx\omega}{(N - 2)\gamma} \left( (N - 1) \left( \frac{1}{N} \frac{\gamma}{2x\omega} \right)^{(N-2)/(N-1)} - 1 \right)$$

$$1 = \frac{2N\omega}{(N - 2)} \left( (N - 1) \left( \frac{1}{2Nx\omega} \right)^{(N-2)/(N-1)} - 1 \right).$$

$$2N\omega \frac{1}{N - 1} = x \frac{(N - 2)}{N - 1} N(1 + 2\omega) - 2.$$  \hspace{1cm} (7)