When games meet reality: is Zynga overvalued?

Zalán Forró^{*}, Peter Cauwels[†], and Didier Sornette[‡]

Department of Management, Technology and Economics, ETH Zürich

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^{*}zforro@ethz.ch, +41 44 632 09 28

[†]pcauwels@ethz.ch, +41 44 632 27 43

 $^{^{\}ddagger} dsornette@ethz.ch, +41 44 632 89 17$

Abstract

On December 16th, 2011, Zynga, the well-known social game developing company went public. This event followed other recent IPOs in the world of social networking companies, such as Groupon or Linkedin among others. With a valuation close to 7 billion USD at the time when it went public, Zynga became one of the biggest web IPOs since Google. This recent enthusiasm for social networking companies raises the question whether they are overvalued. Indeed, during the few months since its IPO, Zynga showed significant variability, its market capitalization going from 5.6 to 10.2 billion USD, hinting at a possible irrational behavior from the market. To bring substance to the debate, we propose a two-tiered approach to compute the intrinsic value of Zynga. First, we introduce a new model to forecast its user base, based on the individual dynamics of its major games. Next, we model the revenues per user using a logistic function, a standard model for growth in competition. This allows us to bracket the valuation of Zynga using three different scenarios: 3.4, 4.0 and 4.8 billion USD in the base case, high growth and extreme growth scenario respectively. This suggests that Zynga has been overpriced ever since its IPO. Finally, we propose an investment strategy, which is based on our diagnostic of a bubble for Zynga and how this herding / bubbly sentiment can be expected to play together with two important coming events (the quarterly financial result announcement around April 26th, 2012 followed by the end of a first lock-up period around April 30th, 2012). On the long term, our analysis indicates that Zynga's price should decrease significantly.

Keywords: Zynga, valuation, social-networks, IPO, growth in competition, bubble, Facebook, lock-up.

1 Introduction

After the recent initial public offerings (IPOs) of some of the major social networking companies such as Groupon, Linkedin or Pandora, Zynga went public on December 16th, 2011. In November 2011, the estimated value of this social network game developing company was as high as 14 billion USD (Reuters, 2011). However, after the underperformance of the IPO market, this number was scaled down. Indeed, 100 million shares of Class A common stock were sold at 10^{\$} per share, the top end of the indicative 8.5^{\$} – 10^{\$} range (S1/A Form of the Filings to the SEC (a), 2011). With a total of 699 million shares outstanding, the market capitalization of the company at IPO was of 7 billion USD. After dropping to 5.6 billion USD on January 9th, 2012, its minimum since the IPO, and peeking at 10.2 billion USD on February 14th, 2012, the market capitalization of Zynga was around 9 billion USD on February 26th (Bloomberg). The efficient-market hypothesis suggests that such large price changes should reflect significant variations in the fundamentals of the company that lead to the re-assessment of its value by investors and analysts (Fama, 1970). As such, one could question the economic justification for such a change in price in only a few months time.

In addition, during the IPO process, no specifics were given on the methodology to come to Zynga's valuation; neither by the underwriters of the IPO, nor in the S1 filing, nor by the media, nor by investment banking sell-side analysts. The aim of this paper is to determine Zynga's fundamental value and put its current valuation into perspective. For this purpose, we extend to Zynga the methodology proposed by Cauwels and Sornette (2012) for the valuation of Facebook and Groupon, by introducing a semi-bootstrap approach to forecast Zynga's user base.

The pricing of IPOs, and companies in general, has been extensively studied. Ibbotson and Ritter (1995); Ritter and Welch (2002), among others, reviewed well-known stylized facts when companies go public. We can cite the underpricing of new issues, i.e. the fact that underwriters underprice the IPO leading to high returns on the first day of trading, or the long-term underperformance of the underpriced IPOs compared to their "fairly" priced counterparts. During the dot-com bubble, the rapid rise of the Internet sector contrasting with the modest growth of the "old economy" raised a lot of interest. Bartov et al. (2002) showed that there were differences in the valuation of Internet and non-Internet firms. Notably, for the latter, profits were rewarded (positively correlated with the share value) and losses were not (as is usually the case). However, the reverse was true for Internet companies, where losses were rewarded and profits were not. This somewhat paradoxical situation arose from the perception that losses were not the result of poor company management but rather investments that would later pay off. Demers and Lev (2001); Hand (2001) further showed that webtraffic was an important factor in the market value of the Internet company. Indeed, in the case of web-traffic intensive companies, while losses were being rewarded before the peak of the bubble and profits were not, the situation reversed after the peak of the bubble: profits became rewarded and losses not anymore. This phenomenon was not observed for Internet

companies without web-traffic.

We should notice that the studies, mentioned so far, tried to explain the market price of companies using different explanatory variables (such as revenues, type of company, amount of web-traffic, difference between IPO price and first day closing price and so on), making the implicit assumption that the market is efficient and reflects the intrinsic value of the company. While in the long-run this may be a good approximation, this is not true for shorter time-scales (during a bubble typically). As such, these methods (often based on linear relationships between the market price and the explanatory variables) are not meant to reveal the fundamental value of a company or make long-term predictions.

Ofek and Richardson (2002) tackled the problem from a different angle. They assumed that, on the long-run, the price-to-earnings ratio of the Internet companies would converge to their "old economy" counterparts, and computed the growth in earnings necessary to achieve that. They found unrealistic growth rates making an argument against market rationality. Schwartz and Moon (2000) used a real-option approach to value Amazon, the company having the option to go bankrupt (thus limiting their losses). Their model relies on the future growth rate of revenues and use the discounted cash flows method. It has the upside to come up with a valuation for the company but is very sensitive to variations in its parameters. Gupta et al. (2004) extended a methodology developed by Kim et al. (1995) to value Amazon, Ameritrade, E-bay and E*Trade (Internet companies). Their model uses the discounted cash flows analysis where the future revenues are computed based on the prediction of the company's user base combined with an estimation of the revenues generated by each user in time. They obtain robust valuations of the Internet companies, allowing for a quantitative assessment of the discrepancy between the market capitalization of the companies and their fundamental values. Adopting a similar approach, Cauwels and Sornette (2012) show that Facebook and Groupon are overvalued. The main insight of the aforementioned works is to recognize that, for companies deriving their value directly from their users, such a simple approach can give much better estimate of the intrinsic value of a company than methods employed so far.

The present paper adds to the existing literature by extending the methodology of Cauwels and Sornette (2012) to Zynga, a company where the user dynamics are very different and more complicated from the ones observed in Facebook, Groupon, Amazon and so on. Indeed, the evolution of Zynga's user base is a result of the individual dynamics of each of its individual games and cannot therefore be modeled by a single function. Moreover, we find that the revenues per user have entered a saturation phase. This limits Zynga's ability to increase its revenues much further, as their user base is already in a quasi-stationary phase. Finally, we find that Zynga has been greatly overvalued since its IPO and give a short time scale prediction about its price dynamics by combining our fundamental analysis with the effect consisting of a price drop subsequent to the end of the lock-up period.

This paper is organized as follows. Section 2 gives a brief summary of the methodology used

to value Zynga. Section 3 describes the dynamics of the number of daily active users (DAU) of Zynga. Section 4 analyzes the financial data relevant to the valuation of the company. Section 5 gives its estimated market capitalization. Section 6 analyzes the evolution of Zynga so far in the light of its valuation. Section 7 discusses possible strategies to arbitrage the over-valued stock of Zynga and section 8 concludes.

2 Valuation methodology

The major part of the revenues of a social networking company is inherently linked to its user base. The more users it has, the more income it can generate through advertising. From this premise, the basic idea of the method proposed by Cauwels and Sornette (2012) is to separate the problem into 3 parts:

1. First, we will forecast Zynga's user base. This is what we call the part of the analysis based on hard data and modeling. Because Zynga uses Facebook as a platform, and one does not need to register to have access to its games (a facebook account is sufficient), there is no such thing as a measure of total registered users (U). These registered users were used by Cauwels and Sornette (2012) in their valuation of Facebook and Groupon. Because it takes an effort to unregister, the number of registered users is an almost monotonically increasing quantity. As such, Cauwels and Sornette (2012) were able to model and forecast Facebook and Groupon's user dynamics with a logistic growth model (equation 1).

$$\frac{dU}{dt} = gU(1 - \frac{U}{K}) \tag{1}$$

Here, g is the constant growth rate and K is the carrying capacity (this is the biggest possible number of users). This is a standard model for growth in competition. When $U \ll K$, U grows exponentially since $\frac{dU}{dt} \approx gU$ (this is the unlimited growth paradigm) until reaching saturation when U = K (and $\frac{dU}{dt} = 0$). This model is a good description of what happens in most social networks: the number of users starts growing exponentially and eventually saturates because of competition/constrained environment. For Zynga, a different approach had to be worked out. Here, the analysis is based on the number of Daily Active Users (DAU), a more dynamical measure. DAU can fluctuate (up and down) and as such cannot be modeled with a logistic function. Moreover, Zynga's users form an aggregate of over 60 different games. Therefore, to understand the dynamics of Zynga's user base, we had to examine the user dynamics of its individual games. Figure 1 gives the total number of Zynga users and the DAU of two of its most popular games. We decided to model each of its top 20 games individually, this approach accounting for more than 98% of the recent total number of Zynga users. The specifics of this analysis will be further elaborated in section 3.

- 2. The second part of the methodology is based on what we consider as "soft" data: this part uses the financial data available in the S1/A Filing to the SEC (2011). These will be used to estimate the revenues that are generated per daily active user in a certain time period. It also reveals information on the profitability of the company. Due to the limited amount of published financial information, we will have to rely on our intuition and good-sense to give our best estimate of the future revenues per user generated by the company. This is why we call this part the soft data part. It will be further elaborated in section 4.
- 3. The third part combines the two previous parts to value the company. With an estimate of the future daily number of users (DAU) and of the revenues each of them will generate (r), it is possible to compute the future revenues of the company. These are converted into profits using a best-estimate profit margin (p_{margin}) , and are discounted using an appropriate risk-adjusted return d. The net present value of the company is then the sum of the discounted future profits (or cash flows).

$$Valuation = \sum_{t=1}^{end} \frac{r(t) \cdot DAU(t) \cdot p_{\text{margin}}}{(1+d)^t} = \sum_{t=1}^{end} \frac{\text{profits}(t)}{(1+d)^t}$$
(2)

Hereby, we optimistically assume that all profits are distributed to the shareholders.

3 Hard Data

3.1 General approach

We will take the following steps to forecast Zynga's DAU:

- 1. We will use a functional form to the DAU of each of the top 20 games to forecast the future DAU evolution of the company. This is done as follows:
 - The data that are available are used as it is.
 - This is extended into the future by extrapolating the DAU-decay process with an appropriate tail function.
- 2. Because Zynga relies on the creation of new games in order to maintain or even increase its user base, it is important to quantify its rate of innovation. This is done by using $p(\Delta t)$ the probability distribution of the time between the implementation of 2 consecutive new games (restricted to the top 20).
- 3. Finally, a future scenario is simulated as follows: for the next 20 years, each Δt days, Δt being a random variable taken from $p(\Delta t)$, a game is randomly chosen from our pool of top 20 games. The DAU of Zynga over time is then simply the sum of the simulated games. A thousand different scenarios are computed.

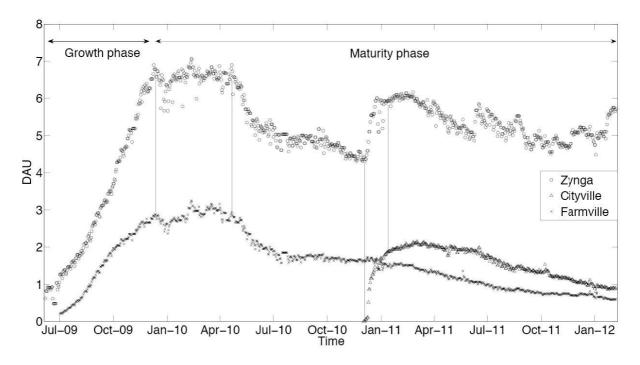


Figure 1: The number of DAU as a function of time for Zynga and two of its most popular games, Cityville and Farmville. After an initial growth period, Zynga entered a quasi-stable maturity phase since January 2010. A typical feature of the games can be seen in Cityville and Farmville: after an initial rapid rise, the DAU of the games enters a slower decay phase. The black vertical lines show that the total DAU of Zynga depends stronly on the performance of the underlying games. Notice that, even though Zynga exists since mid-2007, we do not have DAU data since its beginning. (Source of the data: http://www.appdata.com/devs/10-zynga)

3.2 The tails of the DAU decay process

The functional form of the DAU of each game is composed of the actual observed data and a tail that simulates the future decay process. We will use a power law, $f(t) \propto t^{-\gamma}$, for that purpose. This results in a slow decay process and as such will not give rise to any unnecessary devaluation of the company by underestimating its future user base. Figure 2 shows the power law fits (left) and the extension of the user dynamics into the future (right) for the games Farmville and Mafia Wars.

Such power law is a reasonable prior, given the large evidence of such time dependence in many human activities (Sornette, 2005), which includes the rate of book sales (Deschatres and Sornette, 2005; Sornette et al., 2004), the dynamics of video views on YouTube (Crane and Sornette, 2008), the dynamics of visitations of major news portal (Dezso et al., 2006), the decay of popularity of Internet blogs posts (Leskovec et al., 2007), the rate of donations following the tsunami that occurred on December 26, 2004 (Crane et al., 2010) and so on.

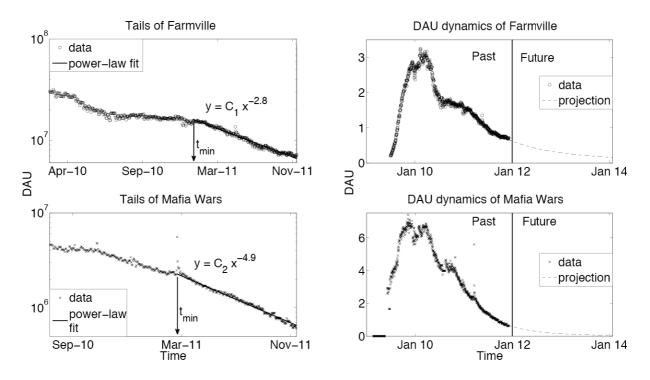


Figure 2: Left: Decay of Farmville (circles) and Mafia Wars (crosses), 2 representative games out of Zynga's top 20. It can be seen that a power law is a good fit for the tails from t_{min} onwards. Right: Simulated dynamics of the games based on the power law parameters of the left panel. (Source of the data: http://www.appdata.com/devs/10-zynga)

3.3 Innovating process

To be able to realistically simulate Zynga's rate of innovation, it is important to understand the generating process underlying the creation of new games. The simplest process that can be used for that purpose is the Poisson process. To understand its meaning, consider the Bernouilli process, its discrete counterpart. It has a very intuitive meaning and can be thought of as follows: at each time step, a game is introduced with a probability of p (and no game is introduced with a probability of 1-p). For a large enough number of time steps and a small enough p, the Bernoulli process converges to the Poisson process. The Poisson process has 3 important properties:

- 1. It has a constant innovation rate.
- 2. It has independent inter-event durations.
- 3. The inter-event durations have an exponential distribution: $p(\Delta t) = e^{-\lambda \Delta t}$.

To assess whether this is a suitable process to model the innovation rate, we measured the time between the introduction of two consecutive new games, $\Delta t_{(1,2)}, \Delta t_{(2,3)}, ..., \Delta t_{(n-1,n)}$, and tested for the above mentioned 3 properties.

3.3.1 Innovation rate

To test whether the innovation rate is constant, different approaches can be adopted. One possibility is to test for the stationarity of the DAU of Zynga since it entered its maturation phase. Stationarity in the number of users would imply a constant innovation rate. Indeed, figure 1 suggests that the user dynamics of Zynga are stationary, its number of DAU being comprised between 43 and 70 million users since the end of the growth phase. However, due to the short time-span of the data, it is hard to implement rigorous statistical tests such as unit-root tests. Instead, we adopt a different approach. If the rate of creation of new games is constant, then the number of new games created as a function of time should lie around a straight line with slope λ , the intensity of the Poisson process. Figure 3 shows the counting of new games as a function of time.

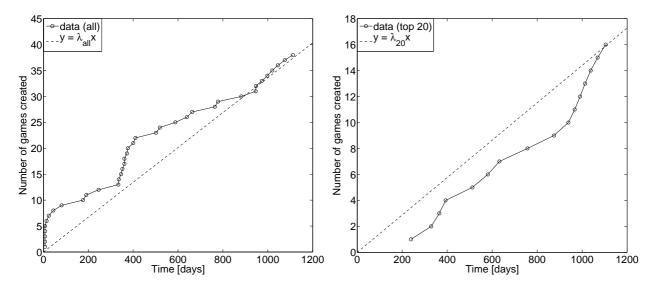


Figure 3: Left: number of newly created games as a function of time for all the games. The empirical innovation rate is at most equal to the theoretical rate coming from the Poisson process (dashed line). The parameter λ_{all} is obtained by using maximum likelihood (assuming a Poisson process). Right: number of newly created games as a function of time for the top 20. The empirical innovation rate seems to be higher for the last games than λ_{20} , the theoretical rate from the Poisson process. This is most likely due to insufficient statistics, this phenomenon being absent when all games are taken into account.

As we can see from figure 3, the constant innovation rate is a good approximation. Our main concern was to discard the possibility of an important increase in the frequency of creation of new games towards the end of the time period, which would have lead to an underestimation of the number of new games created in the future and hence of the future number of users. When all games are taken into account, the innovation rate is at most equal to the one coming from the Poisson process. As such, the Poisson process with constant intensity would not lead to an underestimation of the value of the company.

3.3.2 Independence of inter-event times

To test for the independence of the measured Δt , we study the autocorrelation function $C(\tau)$, the correlation between $\Delta t_{(i,i+1)}$ and $\Delta t_{(i+\tau,i+\tau+1)}$. The result is given in figure 4.

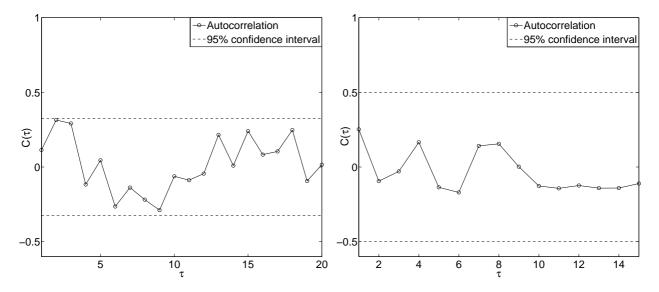


Figure 4: Left: Autocorrelation function for the inter-event times of all the games. The confidence interval (CI) indicates the critical correlation needed to reject the hypothesis that the inter-event times are independent. It is computed as $CI_{.95} = \pm \frac{2}{\sqrt{N}}$, N being the sample size (Chatfield, 2004). Right: Autocorrelation function for the inter-event time of the top 20 games. In both cases, the independence hypothesis cannot be rejected.

As can be seen, $C(\tau) = 0$ is within the confidence interval for $\tau > 0$ in both cases, so the independence hypothesis cannot be rejected.

3.3.3 Distribution of inter-event times

To test for the distribution of inter-event times, we use a Q-Q plot. The Q-Q plot is a graphical method for comparing two distributions by plotting their quantiles against each other. If the obtained pattern lies on a straight line, the distributions are equal. In our case, the two distributions to be compared are the empirical one (from data) and the theoretical one, an exponential with parameters obtained from a maximum likelihood fit to the data. The results of this analysis are presented in figure 5.

As we can see, in both cases the Q-Q plots show a reasonable agreement between the empirical and the exponential distribution given the number of data points, so that the exponential distribution for the inter-event times (Δt) cannot be rejected. The innovation process will thus be modeled as a Poisson process.

3.4 Predicting the future DAU of Zynga

Starting from the present up to the next 20 years, a top 20 game is randomly sampled each Δt days, with Δt drawn from its theoretical exponential distribution $p(\Delta t)$. For each of

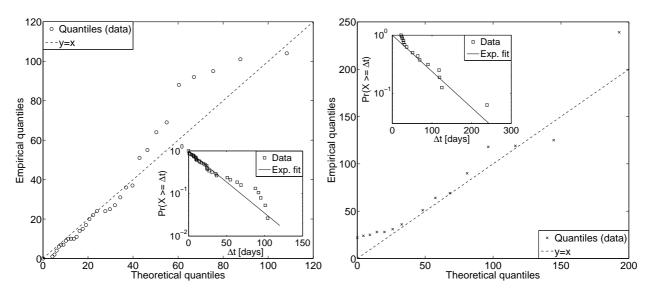


Figure 5: Left: Q-Q plot of the distribution of time intervals Δt between the introduction of new games, for all the games. The theoretical (assuming a Poisson process) and data quantiles (red circles) agree well as can be seen from the proximity to the dashed black line (y=x). The subpanel shows the CDF of the data (red squares) and the exponential fit (black line) on which the quantiles were built. Right: Q-Q plot of Δt for the top 20 games. We can see a deviation for small values of Δt between exponential theoretical and data quantiles. This can be attributed to insufficient statistics.

these sampled games, the DAU is calculated using its functional form. Summing the DAU of all these games, the user's dynamics of Zynga is computed. This process is repeated a thousand times, giving a thousand different scenarios. As can be seen from figure 6, the evolution of the user base between scenarios can be quite different. That is the reason why a wide range of scenarios are needed. The valuation of the company will be computed for each of those scenarios (using equation 2). This will give a probabilistic forecast of the market capitalization of Zynga.

4 Soft Data

The next step to calculate Zynga's value is to estimate the revenues per DAU per year. We base our analysis on the S1/A Form (2011) complemented with the 8-K Form of Filings to the SEC (2012), to add the last quarter of 2011 results. The yearly revenues are given each quarter as a running sum of the four previous quarterly revenues:

$$R_i = R_{i-3}^q + R_{i-2}^q + R_{i-1}^q + R_i^q . aga{3}$$

Here, R_i and R_i^q are respectively the yearly and quarterly revenues at quarter *i* with $i \in (4, \text{last})$. The yearly revenues per DAU at each quarter, r_i , are then obtained by dividing R_i by $\langle DAU_i \rangle_{year}$, the realized DAU at time *i* averaged over the preceding year. Figure 7 gives the historical evolution of the revenues per DAU. Initially, this followed an

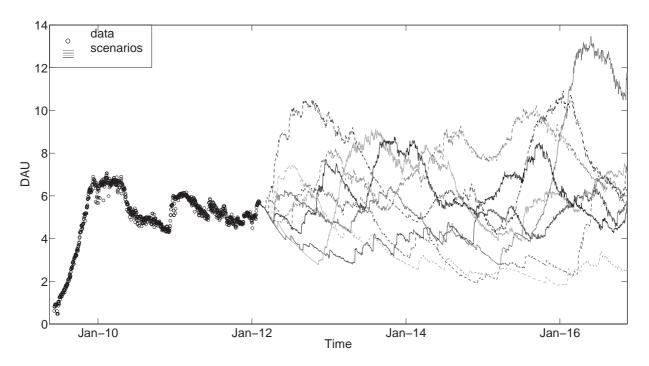


Figure 6: 8 different scenarios of the DAU evolution of Zynga for the next 4 years. The company will be valued for each of these scenarios (section 5). This will give a range for its expected market capitalization.

exponential growth process. However, as can be clearly seen in the right panel, this growth is saturating and the process is following the trajectory of a logistic function. This implies that the revenues per DAU will reach a ceiling. As such, different logistic functions are fit to the dataset. Each of these corresponds to a different scenario: a base case, a high growth and an extreme growth scenario (as defined in Cauwels and Sornette, 2012). They can be seen on the left panel.

From the beginning of Zynga up to April of 2011, both the exponential and the logistic fits perform similarly. Indeed, when $r_i \ll K$, when the revenues per user are far away from saturation, the logistic function can be approximated by an exponential. April 2011 is a turning point in the sense that the growth of the revenues per user slows down, hence the deviation from the exponential (growth at constant rate). This saturation in r_i is easy to explain: there have to be constraints on how much money can be extracted from a user. Under spatial constraints (there is a limited number of advertisements that can be displayed on a webpage), time constraints (there are only so many advertisements that can be shown per day) and ultimately the economic constraints (there is only so much money a user can spend on games or an advertiser is willing to spend), the revenues per DAU are bound to saturate. Using this logistic description for the revenues per DAU, a valuation of Zynga will be given for each of the 3 growth hypotheses.

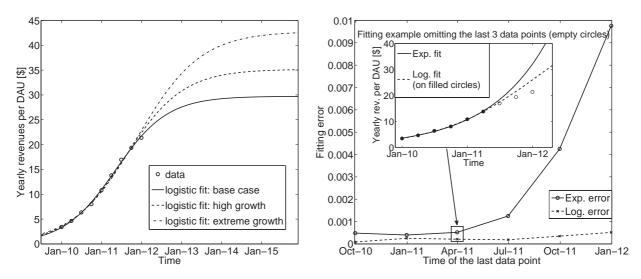


Figure 7: Left: Yearly revenue per DAU over time. A logistic fit (equation 1) is proposed with $K \approx 30, 35$ and 43 USD for the base case, high growth and extreme growth scenarios. Right: Fitting error of the exponential vs logistic function. Each point is obtained by performing the logistic and exponential regressions on data taking more and more data points into account, starting from a minimum of 4 (as shown in the subpanel where only filled circles are fitted). We can see that the logistic starts performing significantly better than the exponential from July 2011. (Source of the data: S1/A and 8-K Forms of the Filings to the SEC.)

5 Valuation

Combining, through equation 2, the hard part of the analysis with the soft part, ie, the number of users over time and the revenues each of them generates per year, the value of the company can be calculated.

We will use a profit margin of 15%. This is Zynga's profit margin of fiscal year 2010. As can be seen from table 1, this is an optimistic assumption since it was the highest profit margin until now, 2010 being the only profitable year of Zynga so far. We also assume that all profits will be distributed to the shareholders and use a discount factor of 5% as in Cauwels and Sornette (2012). We computed the company's valuation for all 1000 different scenarios using equation 2. The results are shown in figure 8 and table 2.

Year	2008	2009	2010	2011
Revenue (millions USD)	19.41	121.47	597.46	1065.65
Net income (millions USD)	-22.12	-52.82	90.60	-404.32
Profit margin	-114%	-43%	15%	-38%

Table 1: Revenue, net	income and	l profit	margin	of	Zynga.	(Source o	f data:	S1/A	and 8	3-K
forms of the filings to	the SEC.)									

We obtain a valuation of 3.4 billion USD for our base case scenario, well below the ≈ 7 billion USD value at IPO or the 9 billion value today (March 2012). Even the unlikely extreme growth case scenario could not justify any of the valuations we have seen in the

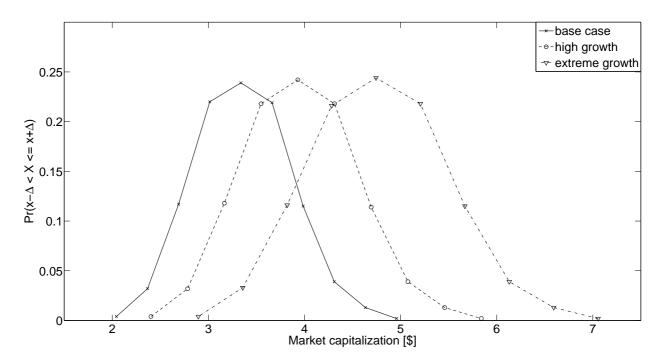


Figure 8: Distribution of the market capitalization of Zynga according to the base case, high growth and extreme growth scenarios. This shows that the 7 billion USD valuation at IPO or today's 9 billion valuation (March 2012) is not even satisfied in the extreme revenues case.

Scenario	Valuation [\$]	95% conf. interval	Share [\$]	95% conf. interval
Base case	3.4 billion	[2.4 billion; 4.4 billion]	4.8	[3.5;6.2]
High growth	4.0 billion	[2.9 billion; 5.1 billion]	5.7	[4.1;7.3]
Extreme growth	4.8 billion	[3.5 billion; 6.2 billion]	6.8	[4.9; 8.9]

Table 2: Valuation and share value of Zynga in the base case, high growth and extreme growth scenarios.

market so far.

6 Historic evolution

At the time of the IPO, on December 15th, 2011, Zynga was valued at 7 billion USD. Right after the IPO, on December 27th, we published an article on Arxiv (Forró et al., 2011) pointing to an overvaluation of Zynga (which was estimated at 4.2 billion USD in our base case scenario). Since then, Zynga published its earnings for the 4th quarter of 2011. These figures increased the accuracy of our valuation since they contributed to reduce the difference between the three scenarios for the revenues per user. By now, it has traded on the stock market for almost 3 months. The big question is whether the share price of Zynga moved into the direction of its fundamental value. As we can see in figure 9, it was quite the contrary: after an initial depreciation of the share value reaching a minimum of 7.97 USD on January 9^{th} (still above our extreme case scenario), it was followed by a moderate run-up in price until February 1^{st} 2012, the date of the S1 filing from Facebook. After that, without any solid economic justification, Zynga skyrocketed to a maximum of 14.55 USD/share corresponding to a 10.2 billion USD valuation on February 14^{th} . However, after the release of the 4th quarter results, the company lost more than 15% in a single day, regaining a part of this loss on the following days.

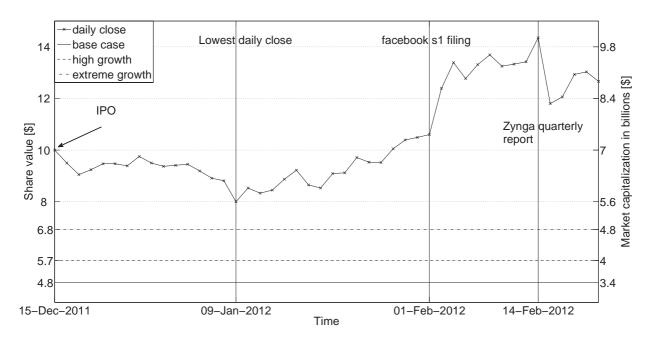


Figure 9: Share value of Zynga over time. The base case, high and extreme growth scenarios are represented (horizontal lines) as well as important dates for the stock price (vertical black lines). (Source of the data: Yahoo Finance)

The highly volatile, news driven behavior of Zynga's stock price can be quantified using the implied volatility measure. In option pricing, the value of an option depends among others on the volatility of the underlying asset. Knowing the price at which an option is traded, one can reverse engineer the implied volatility, the volatility needed to obtain the market value of the option given a pricing model. This standard measure has the upside of being forward-looking: contrary to the historic volatility, the implied volatility is not computed from past known returns. As such, implied volatility is a good proxy for the mindset of the market. Figure 10 compares the implied volatility priced by the market for options written on Zynga with that of Google and Apple.

We can observe a big difference between the two groups. While Apple and Google have a standard stable implied volatility, there is much more uncertainty surrounding Zynga in the eyes of the market, given its high and unstable volatility. What this tells us, until now, is that the market players have a hard time putting a value on Zynga. This perception is reinforced by the following event: on February 15th, 2012, the day following the biggest drop

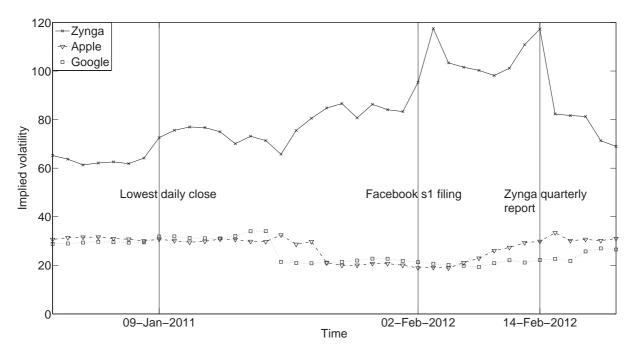


Figure 10: Implied volatility of Zynga, Apple and Google. Zynga has a much higher and less stable volatility than Apple or Google. (Source of the data: Bloomberg)

of Zynga since its IPO, most investment banks (with some exceptions) downgraded Zynga's stock rating, readjusting their price target. Some notable examples of actual price targets per share are (Best Stock Watch, 2012):

- Barclays Capital: 11\$
- BMO Capital Markets 10\$
- Evercore Partners: 10\$
- JP Morgan Chase: 15\$
- Merrill Lynch: 13.5\$
- Sterne Agee 7\$

Compared with our analysis (see table 2), these price targets seem to be high. Moreover, even among "experts", the differences can be significant (the price target of Sterne Agee is less than half of JP Morgan Chase's). One should however keep in mind that the recommendations of most of these companies may not be independent of their own interest, as for example the fact that JP Morgan, Merril Lynch and Barclays Capital are underwriters of the IPO (see Dechow et al. (2000); Michaely and Womack (1999)).

We will have to wait and see how Zynga evolves on longer time scale, but so far, our analysis indicates that Zynga is in a bubble, its price not being reflected by its economic fundamentals.

Zynga may be the symptom of a greater bubble, affecting the social networking companies in general, as suggested by the overpricing of Facebook and Groupon (Cauwels and Sornette, 2012).

7 Arbitraging Zynga's bubble

While the market price of Zynga should converge to its fundamental value on the long run, a prediction of its price movements on a shorter time scale is difficult. It is however possible to develop investment strategies by using a combination of our determination of Zynga's intrinsic value done above with a well known phenomenon, namely the drop of market price when insiders are allowed to sell their shares at the end of their lock-up period.

7.1 End of lock-up and its implication

When a company goes public, only a fraction of their shares are put on the market (14%) in the case of Zynga). The rest of the shares are locked-up for a period of typically 180 days. It is common for IPOs to have a lock-up period in order to prevent insiders from massively selling their shares after the IPO, and thus driving the market value of the company down. There is a vast amount of literature exploring the effect of the end of the lock-up period on the share value of a company. While there is a broad consensus on the fact that companies, on average, experience abnormal negative returns following the end of the lock-up period, different authors give different explanation of this effect. Field and Hanka (2001) find that venture capital backed firms experience the largest price drop at the end of the lock-up period. Bradley et al. (2001) confirm the finding and add that the "quality" of the IPO underwriters as well as the price increase since the IPO is positively correlated with the drop in share value. Gao (2005) finds that firms with the highest forecast bias and the highest forecast dispersion by analysts experience the largest drop. Finally, Ofek (2008) makes the argument that a significant increase in share supply can explain the price drop subsequent to the end of the lock-up period. He further argues that the higher the stock price volatility before the end of the lock-up, the bigger the drop in share value.

While most of the above mentioned authors find that it is difficult to develop an arbitrage strategy to take advantage of this effect, we should stress that they all based their works on samples of companies independent of any view regarding their intrisinc value. If they could bias their sample towards the companies whose market value is significantly higher than their fundamental value, we would expect a different outcome. We hypothesize that the overvaluation of the company would be reflected in its market price, as soon as insiders, better informed about the fundamentals of their company, would be allowed to trade freely, i.e., at the end of the lock-up period. We believe that the information asymmetry between outside traders (the only ones who are allowed to trade the shares of Zynga from its IPO) and insiders would be incorporated in the price formation of such a company, and move its market price towards its fundamental value.

7.2 Prediction for Zynga

7.2.1 Timeline

On March 23^{rd} , Zynga announced, in an S1/A Form of the Filings to the SEC (b), 2012, that inside investors including CEO Mark Pincus would sell about 43 million shares. This move was surprising, since Zynga's inside investors were subjected to a lock-up period ending on May 29th. However, a secondary offering was authorized by the underwriters of the IPO and was concluded on March 28^{th} with the shares being sold over the counter for 12\$ /piece (0.36 \$ of which went to the underwriters). On that day, the share value of Zynga experienced its 4th largest drop, going from 13.02 to 12.24 \$ (this corresponds to a 6% decrease). It should be noted that, subsequent to this transaction, these insiders are again subject to a lock-up period and won't be able to trade until its end. In practice, the remaining locked shares will be released in the market in several steps (see 424B4 Form of the Filings to the SEC, 2012):

- 1. Approximately 115 million shares held by non-executive employees around April 30th (or 3 days after Zynga will disclose its financial statement for the first quarter of 2012).
- 2. Approximately 325 million shares held by non-employee stockholders that have not participated in the secondary offering (see section 7.2) on May 29th, 2012.
- Approximately 50 million shares held by directors, executive employees and the stock holders who participated in the secondary offering (such as Mark Pincus) on July 6th, 2012.
- 4. Approximately 150 million shares held by the same persons as in 3. on August $16^{\rm th}$, 2012.

7.2.2 Effect of the financial results for the first quarter of 2012

When trying to predict the future price movements of Zynga, one cannot ignore the fact that on April 26th, 3 days before the end of the first part of the lock-up period, the company will release its financial results for the first quarter of 2012. To understand the implications of this report on Zynga's price movement, we base ourselves on our diagnostic of a bubble. Indeed, during a bubble, phenomena like herding and imitation are dominant among traders (Sornette, 2003). As such, the market players are very sensitive to new information, giving rise to behaviors inconsistent with its content. We believe that the release of Zynga's financial statement on April 26th can be such an event. According to our model, Zynga's yearly revenue per user is saturating. This can be seen on the left-hand side of figure 7, where the yearly revenue per user, computed at each quarter as the running sum of the four previous quarters, are well fitted by a logistic function. The saturation of Zynga's market valuation. Even with a hypothetical 357 million USD of revenues for the first quarter of 2012 (which will be published on April 26th) meaning a 15 % increase from the 311 million USD of revenues last quarter, the yearly revenues per user would fall right onto our logistic fit. Hence, even a 15

% increase in quarterly revenues would not be sufficient to rationally reject the saturating trend of Zynga's revenues per user that we predict. However, compared with the results of the previous quarter, which saw Zynga's revenue only rise by 1.4% (see figure 11), this would be seen as a very strong performance and would most likely be followed by an increase in share value, even more so in the bubble environment that we diagnose.

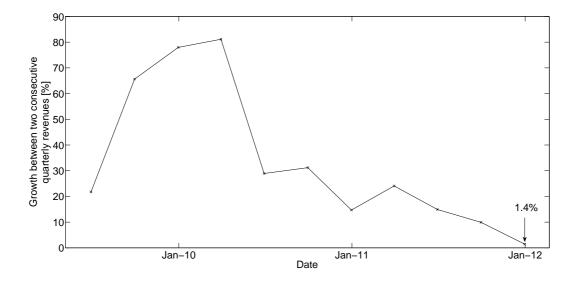


Figure 11: Percentage difference between the revenues of two consecutive quarters. We can see that Zynga's performance in the last quarter was very poor. This suggests that taking the 1.4% figure as a benchmark to evaluate Zynga's performance for the next quarter may lead investors to be overly optimistic, especially during a bubble period (Source of the data: S1 Filing to the SEC).

7.2.3 Effect of the end of the lock-up period

Up to now, there are about 150 million shares tradable on the market (100 million from the IPO and about 50 million from the secondary offering). The 115 million shares coming to the market around April 30th represent an important increase in the free-floating shares of Zynga. As such, and because Zynga satisfies most of the conditions given in subsection 7.1 leading to a large price decrease, we predict a drop of Zynga's market value around that date. We should mention that what happens around April 30th will be conditional on what will have happened on April 26th: we predict this drop to be larger if Zynga's stock price increases on April 26th and smaller if the stock price decreases on the same date. Such a phenomenon could take place at each such dates.

7.2.4 Proposed strategy

We believe that there is a high probability for strong corrections in Zynga's price after each partial lift of the lock-up period. While we have shown that even an apparently strong performance on April 26th would be in line with our diagnostic of Zynga's saturating revenues per user, one should not be surprised to see its share value rise in this bubble environment. On the long-run, we predict Zynga's market value to converge to its intrinsic value of 3.4 billion USD.

In summary, the proposed strategy is based on three time periods:

- From the time of writing (April 16th, 2012) to the announcement of the financial results (around April 26th, 2012): stay out of Zynga or hedge if invested;
- 2. From the day after the announcement (around April 27th, 2012) to end of the first lock-up period (around April 30th, 2012): if the financial results are significantly above those of the previous quarter, buy Zynga for a short term holding period. Otherwise short it.
- 3. From the end of the first lock-up period (after April 30th, 2012): sell the buys of the previous period and short. Monitor the subsequent quarterly releases and the successive ends of future lock-up periods to position a strategy in the same spirit as above.

8 Conclusion

In this paper, we have proposed a new valuation methodology to price Zynga. Our first major result is to model the future evolution of Zynga's DAU using a semi-bootstrap approach that combines the empirical data (for the available time span) with a functional form for the decay process (for the future time span).

The second major result is that the evolution of the revenues per user in time, r_i , shows a slowing of the growth rate, which we modeled with a logistic function. This makes intuitive sense as these r_i should be bounded due to various constraints, the hard constraint being the economic one, since Zynga's players only have a finite wealth. We studied 3 different cases for this upper bound: the most probable one (the base case scenario), an optimistic one (the high growth scenario) and an extremely optimistic one (the extreme growth scenario).

Combining these hard data and soft data revealed a company value in the range of 3.4 billion to 4.8 billion USD (base case and extreme growth scenarios).

On the basis of this result, we can claim with confidence that, at its IPO and ever since, Zynga has been overvalued. Indeed, even the extreme growth scenario (implying 43 USD/DAU at saturation) would not be able to justify any value the company had until now. It is worth mentioning that we adopted a rather optimistic approach:

• We have taken a (slow) power law for the decay process (even in cases where exponentials might be better).

- We chose games only in the top 20 with equal probability in the simulation process (implying that there is the same probability to create a top game and an average/unsuccessful one).
- \bullet We took a 15% profit margin and supposed that all the future profits would be distributed to the shareholders.
- We implicitly assumed that the real interest rates and the equity risk premium stay constant at 0% and 5% respectively, for the next 20 years (Cauwels and Sornette, 2012).

Given these optimistic assumptions, all our estimates should be regarded as an upper bound in our valuation of Zynga.

We should also stress that our assumptions do take into account the innovations that Zynga will have to create in order to continue its business, akin to the Red Queen's race in Lewis Carroll' novel with Alice constantly running but remaining in the same spot.

While the fundamental value computed here suggests that Zynga's market price should decrease on the long term, we were able to delineate an investment strategy built on the expected future price movements of Zynga on a short term scale. In particular, we believe that Zynga's share value will drop significantly around April 30th, 2012 after a possible short run-up following the announcement of the financial results of the first quarter around April 26th, 2012.

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