

## Article

# Production Patterns of Eagle Ford Shale Gas: Decline Curve Analysis Using 1084 Wells

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**Abstract:** This paper analyzes and quantifies characteristic production behavior using historical data from 1084 shale gas wells in the Eagle Ford shale play from 2010 to 2014. Decline curve analysis, using Hyperbolic and Stretched Exponential models, are used to derive average decline rates and other characteristic parameters for shale gas wells. Both Hyperbolic and Stretched Exponential models fit well to aggregated and individual well production data. The hyperbolic model is found to perform slightly better than the Stretched Exponential model in this study. In the Eagle Ford shale play, about 77% of wells reach the peak production of 1644–4932 mil cubic feet per day; the production decline rate of the first year is around 70%, and over the first two years it is around 80%; shale gas wells were estimated to yield estimated ultimate recoverable total resources of 1.41–2.03 billion cubic feet (20 years as life span), which is in line with other studies.

**Keywords:** shale gas; well production; decline curve; Eagle Ford

## 1. Introduction

Shale rocks can be defined as any laminated consolidated rock with >67% clay-sized materials and are typically formed in depositional environments where fine-grained particles fall out of suspension [1]. Significant amounts of organic material can become embedded in shales and allow them to generate hydrocarbons. However, extremely low matrix permeability (typically ranging from nD to  $\mu$ D) is a challenge for commercial exploitation. Extractable natural gas resources trapped in shale formations are called shale gas.

A report by the US Energy Information Administration (EIA), which covered 137 shale formations in 41 countries, estimated that the technically recoverable shale gas resources in the world are 7201 trillion cubic feet (Tcf) [2]. Shale gas deposits have been known and exploited for many decades, although only on a minor scale. It was not considered to be a significant factor until mid-2000s, when wide-spread use of horizontal drilling and hydraulic-fracturing technologies spurred a commercial boom in shale production that is known as the “shale revolution” in the USA [3]. Figure 1 shows the historic and expected production of natural gas by different sources in the USA. This caught the eyes of many scholars, policymakers, and other stakeholders in all over the world and triggered global interest in shale gas potential. Many countries/regions have expressed hopes for developing their domestic unconventional gas resources, particularly shale gas, and are in various stages of planning and evaluation to lay groundwork for potentially larger commercial undertakings in the future. This includes countries such as China, Poland, Ukraine, Mexico, India, Argentina, and Australia, and so forth.

The present global primary energy requirements are met with 85.9% from fossil fuels that contain high percentages of carbon and include petroleum (32.9%), coal (29.2%), and natural gas (23.8%) [4]. Natural gas is relatively cleaner than oil and coal and can partly replace them during their consumption. Hence, it is more and more commonly believed helpful to offer a solution for improving the air quality and mitigating manmade climate change [5]. In addition, considering volatile oil prices and an unstable international oil market, natural gas has an increasingly important role in energy security due to significant untapped resources offering future potential for exploitation. Therefore, many countries have strong demand for natural gas as a crucial pillar in their energy consumption structure due to concerns over environmental protection and energy security [6].

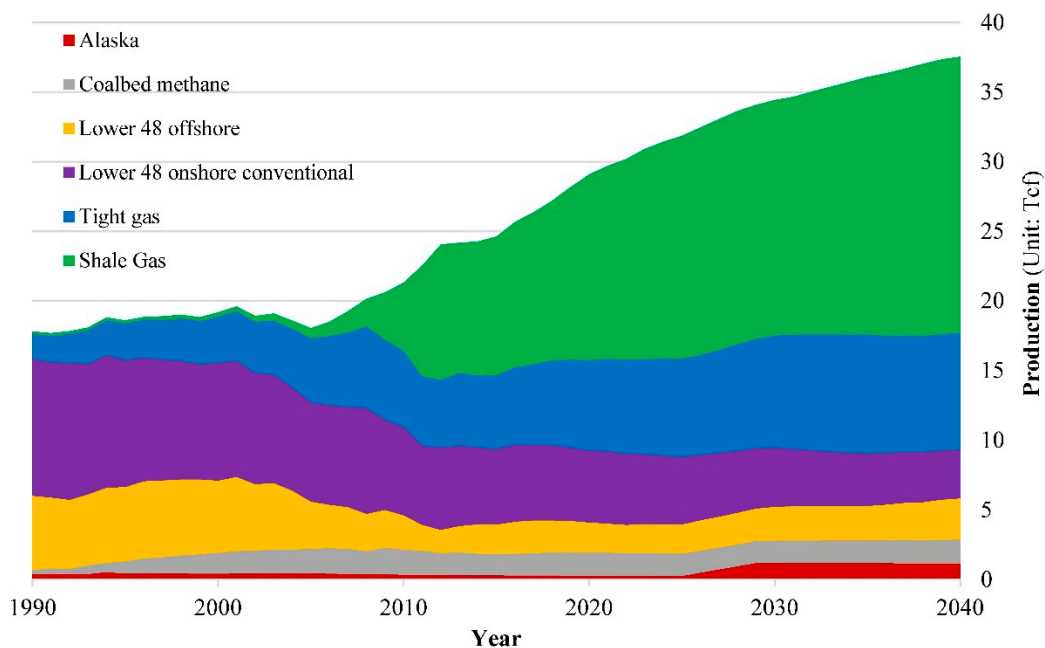


Figure 1. US natural gas production by sources, 1990–2040; Source: EIA 2014 [7].

The rise of the unconventional gas supply is going to influence the natural gas world market and international trade in the following years and reduce dependence on the traditional biggest producers such as Russia, the Middle East, and North African countries [8]. Shale gas, perceived as one of the largest sources of natural gas in the future [9], has a remarkable opportunity to become a more important part of the global energy mix. However, shale gas in regions outside North America is still largely untapped. For example, China has large reserves of shale gas [2] and a pressing demand for natural gas to reduce the consumption of coal, so the Chinese central government has issued policies and development plans to stimulate the domestic shale gas industry. However, the industry is still in the preliminary stage because of many challenges: (1) China's shale geological and surface conditions are more complicated than the U.S., which makes exploration more difficult and increases costs; (2) some key technologies and equipment in China are underdeveloped, also resulting in higher costs and constraints; (3) most of China's shale gas resources are located in densely-populated areas, which may trigger conflicts with local residents due to public health concerns over water contamination, air pollutants, and noise pollution; (4) monopoly on pipeline network, water scarcity, and so on [10]. In addition, there are studies that also show that the process of shale gas development may raise environmental concerns, such as water contamination, air pollution, earthquakes, nuisances and health concerns, and other uncertain impacts [10]. In short, successful exploitation of shale gas resources requires adequate technology, water resources, and evaluation of environmental and social impacts, where many of these still remain unclear [11]. In summary, many knowledge gaps need to be filled to identify the best paths for future development of shale gas resources.

Shale gas has specific production patterns that differ from that of conventional natural gas due to the geological conditions and production characteristics as a low-permeability reservoir. The aim of this study is to identify and compile characteristic production behavior factors for shale gas wells such as decline rate, initial production, and the estimated ultimate recoverable resource (URR) using decline curve analysis. These parameters are useful for both industrial management and operation planning. The data is still very limited, since commercial production of shale gas only has been done on large scales in a few places for barely a decade. Eagle Ford, one of the most successful shale plays in the USA, is used as a case study to provide well-founded empirical data from a large dataset of operating wells.

## 2. Methodology

### 2.1. Decline Curve Models

Depletion occurs when any resource is extracted faster than it can be reproduced by nature [12]. Hydrocarbon resources are only reproduced on geological time scales than require millions of years, while manmade extraction takes years or decades at most. This makes hydrocarbon irreversibly unsustainable for all practical purposes, yet their exploitation is important for many aspects in modern society.

Decline curve analysis is a methodology focused on fitting observed production rates of a single well, or group of wells, by a mathematical function to predict future performance in the future by extrapolating the fitted decline curve function [13,14]. It has been around since 1940s and is used as a benchmark for simple well analysis. The original framework presented by Arps [15] has been proven to be easy to implement and useful for describing and predicting production in conventional oil and gas wells. The decline rate, denoted by  $\lambda$ , can be expressed using derivatives with production rate  $q$  and time  $t$ , as shown in Equation (1):

$$\lambda = -\frac{dq/dt}{q} = Cq^\beta \quad (1)$$

In the Equation,  $C$  is a constant and  $\beta$  is the decline exponent. Arps identifies three types of production rate decline behavior: Exponential (when  $\beta = 0$ , Equation (2)), Hyperbolic (when  $0 < \beta < 1$ , Equation (3)), and Harmonic (when  $\beta = 1$ , Equation (4)):

$$q(t) = q_0 e^{-\lambda(t-t_0)}, \quad (2)$$

$$q(t) = q_0 [1 + \lambda\beta(t - t_0)]^{-1/\beta}, \quad (3)$$

$$q(t) = q_0 [1 + \lambda(t - t_0)]^{-1}, \quad (4)$$

where  $q(t)$  is the production rate at time  $t$ , and  $q_0$  is an initial production rate at time  $t_0$  from which production begins to decline.

The traditional Arps relations were derived for stable reservoir conditions with boundary-dominated flows and should only be used for situations with  $0 < \beta < 1$  [16,17]. Analysts occasionally allow  $\beta > 1$  to obtain better fits, but this extend Arps curves beyond their original region of validity. In transient flow situations, usually lasting for months or even years in shale wells, the validity of hyperbolic decline falters due to very low reservoir permeability and this results in  $\beta > 1$  [18]. This is why  $\beta > 1$  is allowed in the fitting of hyperbolic curves in this study.

To circumvent the limitations of the Arps curves, some new curves have been proposed. Such as the Stretched Exponential (SE) decline model [19], the Power Law (PL) model [16], and Duong's model [20].

Two different decline curve models, the traditional Arps Hyperbolic model and the Stretched Exponential model, are selected and pitted against each other in this study. The advantages of these

decline curves lie in the strong empirical compliance and ease of use with low requirements on input data. Their key properties are described in Table 1. In Section 3.2.3, the accuracy between the two selected models are compared based on the Eagle Ford case.

**Table 1.** Key properties of the Hyperbolic model and the SE model.

Properties	Hyperbolic	SE
$q(t)$	$q_0 [1 + \lambda\beta(t - t_0)]^{-1/\beta}$	$q_i \exp(-D_i t^n)$
$Q(t)$	$Q_0 + \frac{q_0}{\lambda(1-\beta)} \left\{ 1 - [1 + \lambda\beta(t - t_0)^{1-\frac{1}{\beta}}] \right\}$	$Q_0 + \frac{q_i \tau}{n} \left\{ \Gamma\left[\frac{1}{n}\right] - \Gamma\left[\frac{1}{n}, \left(\frac{t}{\tau}\right)^n \right] \right\}$
URR	$Q_0 + [q_0/\lambda(1-\beta)]$	$Q_0 + \frac{q_i \tau}{n} \Gamma\left[\frac{1}{n}\right]$

In the table,  $Q(t)$  is the cumulative production,  $Q_0$  is the initial cumulative production when the decline starts, and URR is the estimated ultimate recoverable resource. In the stretched exponential method,  $q_i$ ,  $D_i$ , and  $n$  are undetermined parameters;  $\tau$  is equivalent to  $(n/D_i)^{1/n}$  [20].

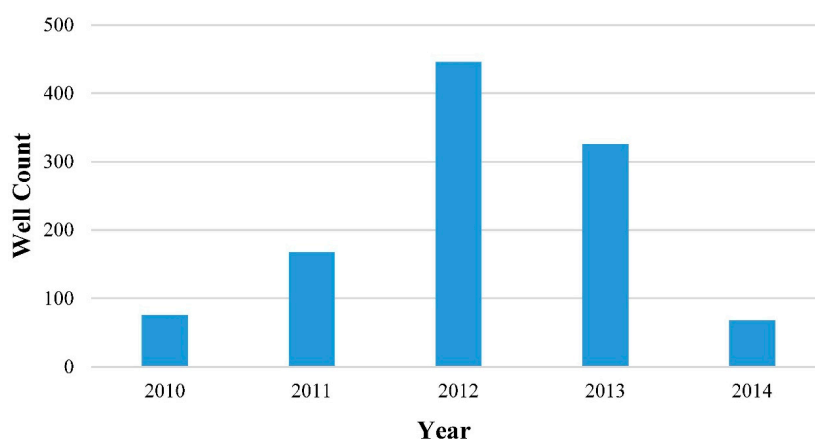
## 2.2. Goodness of Fit

Matlab is used to fit the selected models to actual production data and assess how well the models represent the data. Both the coefficient of determination ( $R^2$ ) and the normalized root mean square error (N-RMSE) are used as measures for the goodness of fit [21,22].  $R^2$  ranges from 0 to 1, and the goodness of fit will become better closer to 1. N-RMSE also ranges from 0 to 1, but here the goodness of fit will become weaker closer to 1. The boundaries that are introduced to exclude the poorest fits are  $R^2 \geq 0.8$  and  $N\text{-RMSE} \leq 0.2$  at the same time.

Moreover, the number of observed data series should be long enough. According to the characteristics of shale industry and the data set, the period of two years (24 months) was set as the minimum number of production month to avoid over-fitting.

## 2.3. Data

The study is based on the monthly well-production data available from DrillingInfo [23], including 1084 shale gas wells that commenced production in January 2010 with data culminating in October 2014 in Eagle Ford. Only horizontal wells whose production is reported separately are included as the data samples used here. When production from several wells is reported together, the temporal distribution of the wells starting points also affects the overall production and this may obfuscate the characteristic production patterns of interest. Figure 2 shows the original number of wells in the data set per year.



**Figure 2.** The original number of wells (1084 in total).

Months with zero production should be removed before analysis, which is a measure to remove external events that affected production such as annual maintenance and scheduled downtime. Curve fitting of the aforementioned models is only done on production data from the decline phase. Figure 3 shows a stylized conceptual production curve with different production phases marked. The peak production point is also the initial production (IP) for the decline phase and one of the most influential factors for the total production. Pre-peak production is minor in comparison to post-peak production. On average, among the investigated wells in the study, the initial cumulative production ( $Q_0$ ) accounts for about 7% of their total well-production ( $Q(t)$ ), and its percentage of URR would be even less, implying that the decline curve captures the vast majority of all production. The relation between  $Q_0$ ,  $Q(t)$ , and URR are also illustrated in Figure 3.

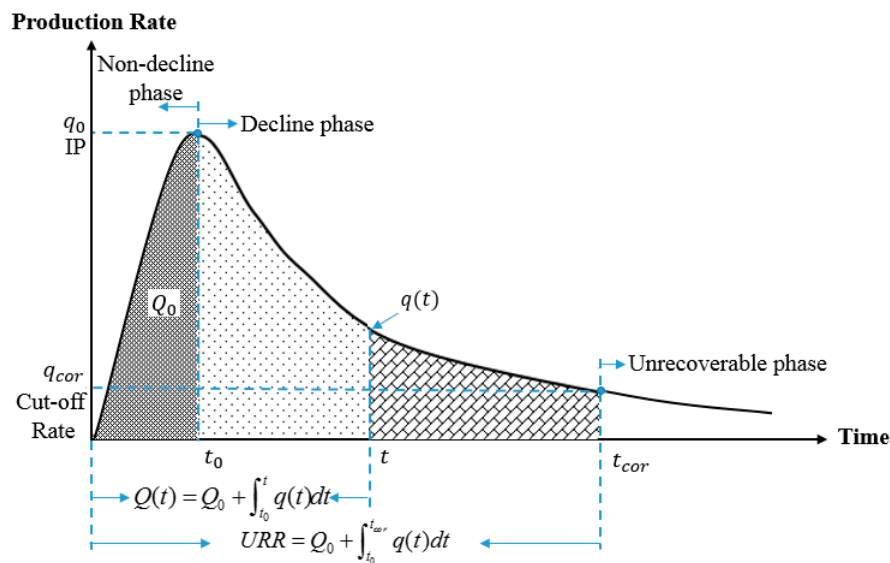


Figure 3. The conceptual production curve.

### 3. Results

#### 3.1. Aggregate Decline Curves

Wells with production data exceeding 24 months are normalized well by well in the sense that the peak production (marked as IP, initial production, in Figure 3) is set to 1 and all following monthly productions are displayed in relation. The aggregated normalized production of the first four years (48 months) and well count used for different months are shown in Figure 4. The tail of the production curve is less certain than the early months as the number of data points shrinks over time. However, it is clear that decline slows down with time. The annual decline over the first year of production is 68.54%, while it is only 43.08% over the second year. Over the first two years, production declines by 82.09% from the initial production level.

Hyperbolic and SE models are fitted to the aggregated normalized production from the Eagle Ford wells. Resulting fits are very similar and hard to distinguish on a regular plot, and for this reason a logarithmic scale is used in Figure 5. Both curves have excellent goodness of fit using  $R^2$  and N-RMSE. For the Hyperbolic model, the  $R^2$  value is 0.9987 and the N-RMSE value is 0.0073. In comparison, the Stretched Exponential model has an  $R^2$  value of 0.9985 and an N-RMSE value of 0.0080. The parameter values of the curves are also shown in the text box in Figure 5.

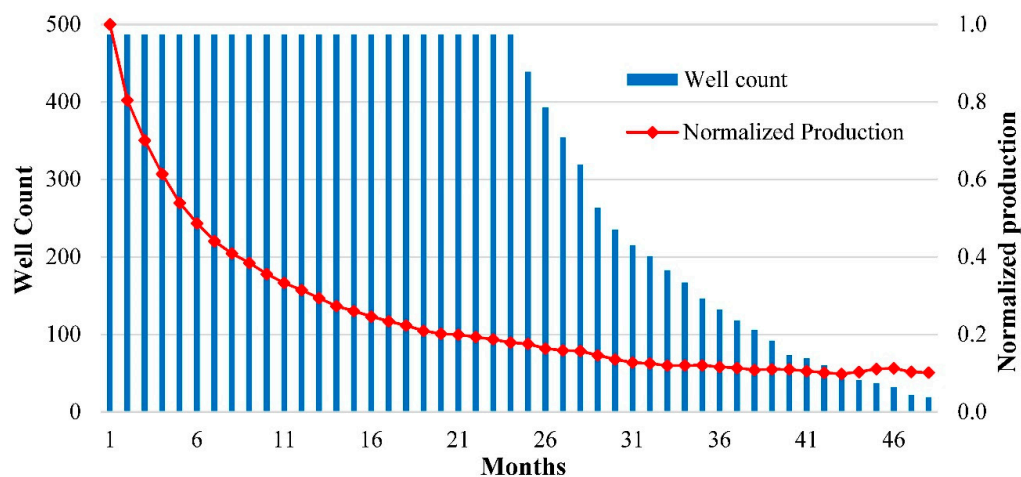


Figure 4. The aggregate decline curve.

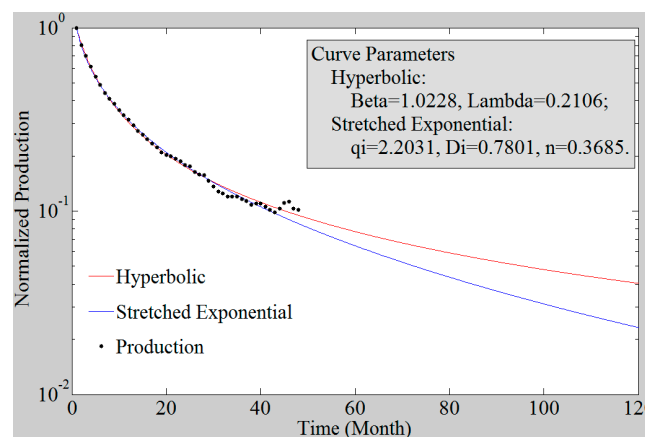


Figure 5. The Hyperbolic and the Stretched Exponential decline curves fitted to average normalized production data on logarithmic scale. Only after 40 months do the two models start to diverge.

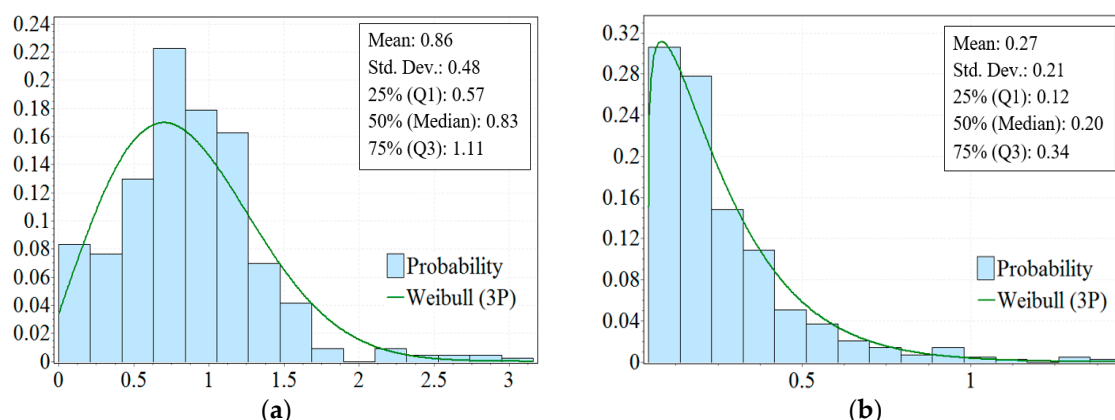
### 3.2. Individual Well Decline Curves

Wells with production data exceeding 24 months are also fitted one by one using both Hyperbolic and Stretched Exponential models. This provides an alternative insight into well behavior compared to aggregating production data prior to curve fitting.

#### 3.2.1. Hyperbolic Decline Curve

The distributions for the best fitted  $\beta$  and  $\lambda$  values of the Hyperbolic fits are displayed in Figure 6a,b respectively with their key descriptive statistics attached in the textboxes. The Weibull distribution (Three parameters: Weibull (3P) probability density function is  $f(x) = \frac{\alpha}{\beta} \left(\frac{x-\gamma}{\beta}\right)^{\alpha-1} \exp\left[-\left(\frac{x-\gamma}{\beta}\right)^\alpha\right]$ , where  $\alpha$ ,  $\beta$ , and  $\gamma$  are fitted parameters.) is one of the best descriptions for the probability distributions according to Chi-Squared tests and this reflects the earlier results found for conventional oil by Höök [24].

The largest value is 3.16 for  $\beta$  and about 37% of the  $\beta$  values are larger than 1, and values between 0.4 and 1.6 make up about 80% of all occurrences (Figure 6).



**Figure 6.** (a) Distributions of  $\beta$  values.  $\beta$  values are on the  $x$ -axis and the probability on the  $y$ -axis, the Weibull parameters are  $\alpha = 2.13$ ,  $\beta = 1.11$ ,  $\gamma = -0.13$ ; (b) Distributions of  $\lambda$  values.  $\lambda$  values are on the  $x$ -axis and the probability on the  $y$ -axis, the Weibull parameters are  $\alpha = 1.15$ ,  $\beta = 0.23$ ,  $\gamma = 0.04$ .

### 3.2.2. Stretched Exponential Decline Curve

Some key descriptive statistics of the best fitted parameters for the Stretched Exponential decline curves are shown in Table 2.

**Table 2.** Descriptive statistics of the SE curves.

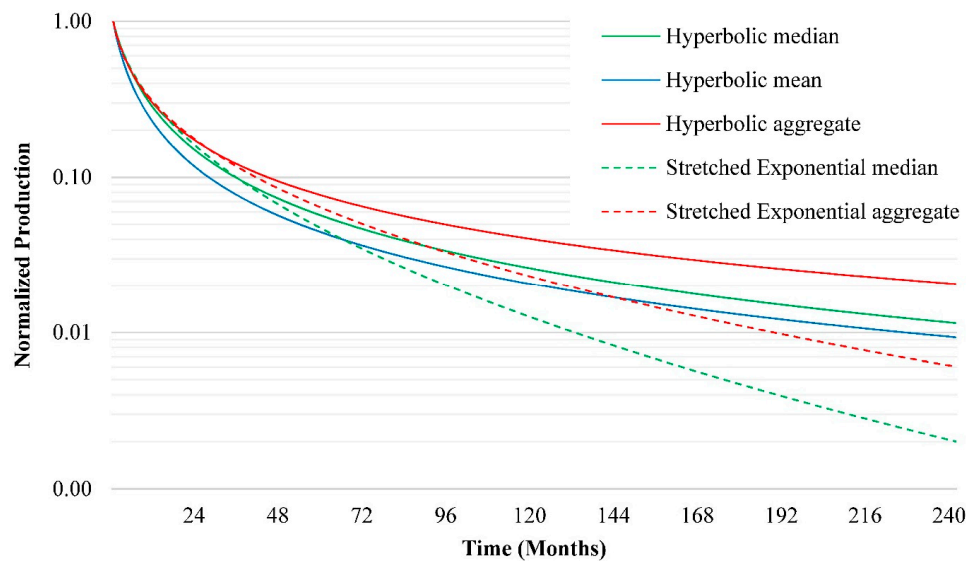
	$q_i$	$D_i$	$n$
Mean	138.95	1.39	0.49
Std. Deviation	589.07	1.95	0.30
Min	0.86	<0.01	0.06
25% ( $Q_1$ )	1.34	0.27	0.28
50% (Median)	1.84	0.58	0.45
75% ( $Q_3$ )	4.26	1.45	0.63
Max	4608.88	8.46	2.78

### 3.2.3. Summary of Decline Curves

The typical decline curves for shale gas production are derived from historical production profiles by using Hyperbolic and SE models. Both models fit aggregated well and individual shale gas well production data. Comparing goodness of fit indicates that Hyperbolic curves are slightly better fits than Stretched Exponential curves for about 65% of the wells according to  $R^2$  and for about 76% of all wells using N-RMSE. However, the result maybe only represents the Eagle Ford case. Longer data series and larger databases are needed to find out which model is better on a greater scale.

Figure 7 provides a summary of different derived characteristic decline curves investigated in this study, including median and mean versions of the two models. However, the parameters of the SE model are interconnected and by taking the mean, the connection will be lost. In Figure 7, the aggregate decline curves (see Figure 5) are also displayed. The aggregate curves (the red solid line and red dotted line) of two models are declining more gradually than their corresponding mean or median lines. The Stretched Exponential median curve (green dotted line) declines steepest in a medium to a long term perspective while the Hyperbolic median and mean curves (green solid line and blue solid line) decline fast initially after which they flatten out.





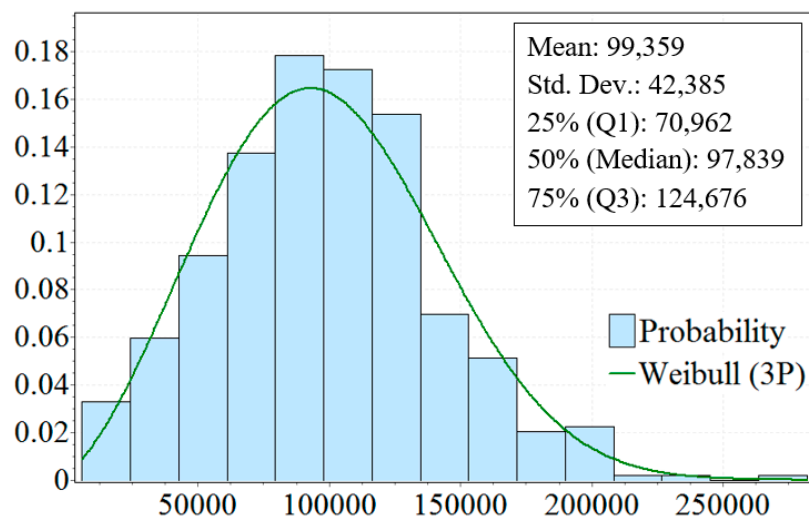
**Figure 7.** Summary of different derived typical decline curves (on logarithmic scale).

### 3.3. Initial Production and Decline Rate

The initial production (IP) at the onset of decline, i.e. peak production, and the decline rates in different time phases of the production curve are also analyzed for all wells with production data exceeding 24 months.

#### 3.3.1. Initial Production

The distributions for the monthly IP for all wells that started production in 2010–2012 are displayed in Figure 8. About 77% of all wells reach a production peak ranging from 50,000 to 150,000 Mcf per month. As shown in the textbox in Figure 8, the mean IP of every well is 99.4 Mcf per month or about 3267 Mcf per day (Mcf/d).

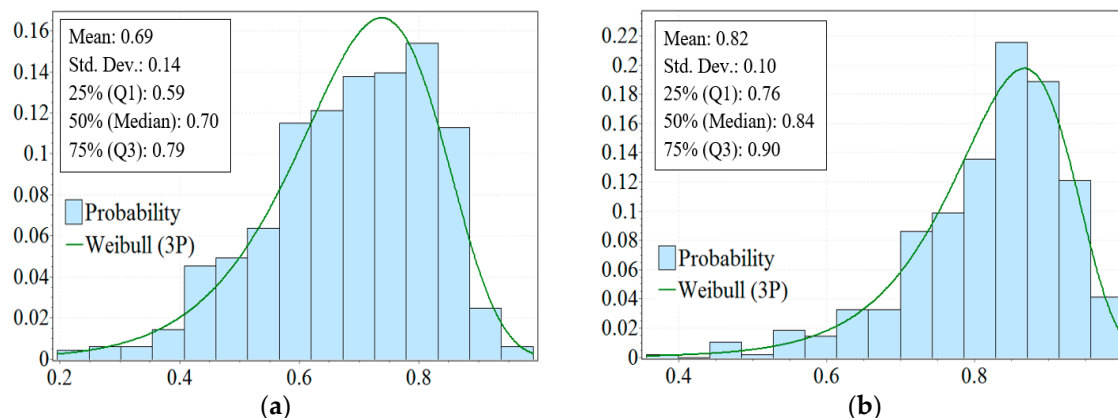


**Figure 8.** Distributions for the monthly IP. Production is on the  $x$ -axis and the probability is on the  $y$ -axis; production unit is Mcf per month, the Weibull parameters are  $\alpha = 2.63$ ,  $\beta = 1.17 \times 10^5$ ,  $\gamma = -5.29 \times 10^4$ .



### 3.3.2. Decline Rates in Different Time Phases

The distributions for the first year (12 months after IP), decline rates, and the decline rates over the first two years (24 months after IP) of every studied wells are displayed in Figure 9a,b, respectively.



**Figure 9.** (a) Distributions for the first year decline rates. Rates are on the  $x$ -axis and the probability is on the  $y$ -axis, the Weibull parameters are  $\alpha = 14.27$ ,  $\beta = 1.68$ ,  $\gamma = -0.93$ ; (b) Distribution for the first two years decline rate. Rates are on the  $x$ -axis and the probability is on the  $y$ -axis, the Weibull parameters are  $\alpha = 8.16 \times 10^7$ ,  $\beta = 6.52 \times 10^6$ ,  $\gamma = -6.52 \times 10^6$ .

The average value of the first year decline rates for all wells started in 2010–2012 is 68.54% and the first two years decline rate is 82.09%; for the annual results, the first year and first two years average decline rates are shown in Table 3.

**Table 3.** Average value of the first year and the first two years decline rates for new wells annually.

	2010	2011	2012	2010–2012
First year	70.58%	70.40%	66.76%	68.54%
First two years	83.72%	83.74%	80.57%	82.09%

### 3.4. Estimated Cumulative Production

At some point, a shale gas well will experience a cut-off rate for continued production (due to technical and economic limits,  $q_{cor}$  in Figure 3), as production flows simply become too low to merit continued operation. Then the ultimate production can be described at the cumulative production over the well's life time to final shutdown, and the estimated  $Q(t)$  at the time where the cut-off rate appear ( $t_{cor}$ ) can be used as an estimate for URR. Kaiser and Yu [25] found that typical wells in Texas are shut down and retired at an average annual production of 5 boe/d. However, the scope of this study does not allow deeper analysis of the economic aspects of shale gas production and projections on future profitability. Instead, assumed life spans are used as an alternative approach.

Applying characteristic decline curves from Figure 8, the mean value of the peak production of 3267 Mcf/d and supposed producing time length of 10-year, 15-year, 20-year, and 30-year are used to estimate expected cumulative production  $Q(t)$ . The average initial cumulative production before the peak is 72,078 Mcf, which is equivalent  $Q_0$  in the  $Q(t)$  functions shown in Table 1. The results are displayed in Table 4.

Production levels would be very low about 10 years after the peak and could be below the economic limit due to characteristic high decline rates. Hence, assumptions of very long life spans (>20 years) for shale wells must be properly grounded. Using a 20-year well life span, shale gas wells were estimated to yield an URR of 1.41 to 2.03 Bcf based on the two decline curve models. Some estimated URR values for typical shale wells found in other studies are summarized in Table 5.

There are significant differences among published estimates due to the short operation history of large-scale shale gas exploitation. For example, mean estimates for URR ranges from 1.0–5.0 Bcf/well in Table 5. Among the studies listed in Table 5, the EIA gives the highest estimate of URR, which is far different from later studies. The main reasons include: (1) the EIA did not estimate the URR and other average properties by itself instead of introducing results reported by some companies such as Petrohawk Energy, Talisman Energy, and Rosetta Resources; (2) the Eagle Ford shale was first discovered by Petrohawk in 2008, so the production history of drilling horizontal wells in the Eagle Ford shale was too short for good URR estimation in 2011. Similar problem also appeared in Baihly's study in 2010. The result of this study is more accurate based on longer and relatively complete data compared with earlier studies. Also, the used goodness-of-fit measures also improve the results. In addition, this study indicates that the lower span of this interval is more reasonable, but more research is encouraged to illuminate this issue further.

**Table 4.** The average expected cumulative production per well.

			10-Year	15-Year	20-Year	30-Year
Hyperbolic	Median	$q(t)$	85	53	38	23
		$Q(t)$	1.49	1.61	1.69	1.80
	Mean	$q(t)$	68	43	31	19
		$Q(t)$	1.25	1.35	1.41	1.50
	Aggregate	$q(t)$	132	90	68	46
		$Q(t)$	1.40	1.89	2.03	2.24
SE	Median	$q(t)$	41	15	7	2
		$Q(t)$	1.44	1.49	1.51	1.52
	Aggregate	$q(t)$	76	36	20	8
		$Q(t)$	1.58	1.68	1.73	1.78

Note: Unit for  $q(t)$  is Mcf/d and for  $Q(t)$  is Bcf (billion cubic feet).

**Table 5.** Summary of average estimated URR per well in Eagle Ford from other studies.

Reference	Mean URR (Bcf)	Publish Time
Gong [26]	1.45–3.78	2013
Hughes [27]	2.36	2013
USGS [28]	1.104	2012
Swindell [29]	1.044	2012
EIA [30]	5.0	2011
Baihly [31]	3.793	2010
This study (20 years as well life span)	1.41–2.03	

#### 4. Conclusions

This study found that, most of all, investigated shale gas wells in the Eagle Ford shale play have a characteristic production pattern characterized by quickly reaching a peak production followed by steep declines (~70% per annum) and low long-term production levels. As shown in the results section, the first annual year decline rate of production of a shale gas well is around 70% and over the first two years about 80% of the initial production level is lost due to decline, which is far higher than that of conventional natural gas. Furthermore, about 77% of wells reached the peak production from 50,000 to 150,000 Mcf per month or 1644 to 4932 Mcf per day. Raising outputs significantly from shale gas wells by enhancing recovery efficiency technology is also an efficient way to counteract this high decline rate.

Sustaining shale gas production in long-term energy strategies requires proper planning to handle the challenges imposed by the characteristic production patterns. A large number of “new wells” are annually needed to offset the production decline from the “old wells” to maintain regional production

levels or increase production output. This is both a logistic challenge in maintaining a host of active drilling rigs, but also a social challenge to obtain acceptance for high drilling activity. Holistic analysis that includes economic balance based on the characteristic shale gas production pattern as well as logistics of the necessary drilling activities are essential to pinpointing the best options for developing a shale play.

According to the results of goodness-of-fit ( $R^2$  and N-RMSE), both the Hyperbolic model and the Stretched Exponential model fits well to aggregated well data and to individual wells. The Hyperbolic model is slightly better than the Stretched Exponential model based on the Eagle Ford case in this study, there are about 37% of the  $\beta$ -parameter values of Hyperbolic larger than 1. However, further research is needed to find out which model is better on a greater scale. On average, shale gas wells were estimated to yield an URR of 1.41 to 2.03 Bcf based on the two decline curve models when 20 years was set as well life span, which is in line with some other studies. As mentioned in the methodology Section 2.1, the traditional Arps model (Hyperbolic) was supposed to be used only for describing and predicting production in conventional oil and gas wells. While this study indicates that it is still effective for a shale gas well if the life span is set reasonably (20 years, for example). Deeper analysis of the economic aspects, such as the cut-off-rate for continued production and further research with more accuracy, are still needed to be able to estimate the future production and resources.

Available production data is still fairly limited for long-term trend analysis, since large scale shale gas operations have only been ongoing for barely a decade. The “American experience” is the only source of empirical data for shale gas production behavior and it serves as a suitable foundation for any attempt to describe or predict future shale gas developments. Production patterns of shale gas wells have general characters, even though the geological conditions for shale plays may vary. The findings of this study provide better understanding of characteristic production behavior and what to expect from shale gas production. The parameters of the models analyzed in the study can be adjusted to describe and predict shale gas production beyond the Eagle Ford shale, and beyond North America. Hence, the findings are relevant for analysts and policymakers to understand and provide access to energy resources required for future development.

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**Author Contributions:** Mikael Höök and Baosheng Zhang conceived and designed the experiments; Keqiang Guo performed the experiments; Keqiang Guo and Mikael Höök analyzed the data; Kjell Aleklett contributed data; Keqiang Guo wrote the paper.

**Conflicts of Interest:** The authors declare no conflict of interest. The founding sponsors had no role in the design of the study; in the collection, analyses, or interpretation of data; in the writing of the manuscript, and in the decision to publish the results.

## Abbreviations

The following abbreviations are used in this manuscript:

SE	Stretched Exponential
Tcf	Trillion cubic feet
Bcf	Billion cubic feet
Mcf	Mil cubic feet = Thousand cubic feet
URR	Ultimate recoverable resource

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