

PERSONALITY & SOCIETY | ISSN: 2712-8024  
Volume 2, Issue 4 – July – December, 2021 / pp 4-10  
Doi: <https://doi.org/10.46502/issn.2712-8024/2021.4.1>

Received: 10.08.2021 | Accepted: 10.09.2021

How to Cite  
Mikhailova, S. V. (2021). "Fakes" in social networking media and modeling "fake infection". *Personality Society*, 2(4), 4-10. <https://doi.org/10.46502/issn.2712-8024/2021.4.1>

## "Fakes" in social networking media and modeling "fake infection"

### "Подделки" в социальных сетях и моделирование "поддельной инфекции"

Mikhailova Svetlana Viktorovna<sup>1</sup>

#### Abstract

The growth of dynamism, the complexity of relationships in social networks requires a systematic approach, the development of mathematical models for forecasting and the identification of fake news in social networks. Otherwise, it is difficult to resist media misinformation, fake news. The problem is urgent, there are more and more opportunities for exchanging "viral" and fake messages in social networks, and we poorly implement monitoring, identifying fake risks. Social networks so far do not allow reliably distinguishing lies from news from aggregator. The purpose of the work is to predict and analyze the system-phase pattern of the spread of fakes in the space of social interactions. "Fakes" are deliberately false, intended for manipulation. In recent years, they are easily distributed in social networks. In the work by methods of the theory of ordinary differential equations, their qualitative analysis, the above problem was fully investigated. The study was conducted under assumptions: remote distributors are not allowed to participate in the transmission of fakes; an adult population susceptible to fakes has a constant birth rate; propagation can occur "vertically," wherein the transmission mechanism is introduced into the model by appropriate assumptions about the proportion of susceptible and distributors. The problem is fully investigated (solvability, unambiguity, phase patterns of stable behavior). The work will be useful in the practical identification and prediction of the influence of fake news.

**Keywords:** fake news, mathematical modeling, social networks, vertical distribution, identification.

#### Аннотация

Рост динамизма, усложнение взаимоотношений в социальных сетях требует системного подхода, разработки математических моделей для прогнозирования и выявления фейковых новостей в социальных сетях. В противном случае трудно противостоять дезинформации СМИ, фальшивым новостям. Проблема актуальна, появляется все больше возможностей для обмена "вирусными" и поддельными сообщениями в социальных сетях, и мы плохо реализуем мониторинг, выявляющий поддельные риски. Социальные сети пока не позволяют надежно отличить ложь от новостей от агрегатора. Целью работы является прогнозирование и анализ системно-фазовой закономерности распространения фейков в пространстве социальных взаимодействий. "Подделки" являются заведомо ложными, предназначенными для манипулирования. В последние годы они легко распространяются в социальных сетях. В работе методами теории обыкновенных дифференциальных уравнений, их качественного анализа была полностью исследована вышеуказанная проблема. Исследование проводилось в предположениях: удаленным распространителям не разрешается участвовать в передаче подделок; взрослое население, восприимчивое к

<sup>1</sup> Assistant. Industrial University of Tyumen, Russia, <https://orcid.org/0000-0003-1292-2750>



поделкам, имеет постоянный уровень рождаемости; распространение может происходить "вертикально", при этом механизм передачи вводится в модель с помощью соответствующих предположений о доле восприимчивых и распространителей. Проблема полностью исследована (разрешимость, однозначность, фазовые закономерности устойчивого поведения). Работа будет полезна при практическом выявлении и прогнозировании влияния фейковых новостей.

**Ключевые слова:** фейковые новости, математическое моделирование, социальные сети, вертикальное распределение, идентификация.

## Introduction

The growth (in volume, dynamism, complexity, connectivity) of relationships in social networks (hereinafter referred to as social networks) requires the development of intelligent systems and mathematical models (forecasting) of fake news in social networks (see, for example, Tretyakov et. al., 2018; Golovatskaya, 2019). Without this, it is difficult to resist digital, media misinformation (Domagoj & Volarević, 2018; Allen et al., 2020). For example, the Google-request "fake news" produces more than a billion pages, and the request "fake news science article" - 361 million.

There is already a science of fake news (Lazer, Baum, Benkler et al., 2018), the problem of forecasting is all the more relevant, and the more possible it is to exchange "viral" and anonymous messages on social networks. Given the complexity, almost poorly implemented social network monitoring, identification of fake risks and actors.

## Theoretical bases

Social media is part of our society, but can we trust them as a source of news, how to distinguish lies from news aggregators?

A lot, especially in social networks, may seem untrue. False information - news, stories or hoax for intentionally misinforming (deceiving) readers. Typically, such stories are created to influence people's opinions, advance the political agenda or cause confusion, and can often be profitable for online publishers. False information can mislead people by issuing as verified news or using names (addresses) similar to authoritative news aggregators (Sukhodolov & Bychkova, 2017).

"Fake news" or "fake" refers to news or other materials that are deliberately false, intended to manipulate the reader. Although the concept of fakes has existed for a long time, in recent years it has become a big problem due to the ease of its distribution in social networks, on online platforms.

It's important to recognize fake news, prevent its distribution, without transmitting it with its social network capabilities. Several social media platforms responded to the increase in the number of fakes by changing their news feeds, marking the news as false (contested) or using other approaches. Google has also made changes to address this issue.

Traditionally, we receive news from reliable sources, from journalists and the media, which are obliged to comply with strict rules. Many receive news from social networks, where it's often difficult to assess whether they deserve trust. Information overload and general lack of understanding by people of media letters also contributed to an increase in the number of fake news. Sites, social networks, blogs can play a big role in increasing fakes.

Social networks allowed anyone to potentially attract a wide audience. False information can be a profitable business, profitable from publishers' advertising, various media channels and viral marketing. The more clicks fake has, the more advertisers, sites earn. For many publishers, social networks are an ideal platform for increasing web traffic.

The spread of fakes leading to violence, unrest, resistance to the authorities, and hostility of groups can be equated with information weapons of mass destruction. After all, only the daily audience of Yandex. News exceeds 6

million readers. Fakes are able to provoke a semblance of a destructive "information terrorist attack" (Kaziev et al, 2017). | 6

Unfortunately, so far there are no relevant measures against fakes, including legal, international cooperation (Vosoughi, Roy & Aral, 2018). Mechanisms are needed to filter "fake news" in the media, especially, abusing their distribution of media giants (Aral & Eckles, 2019). Here, work is also carried out using artificial intelligence systems and expert ones capable of assessing the measure of authenticity of the news distributed. For example, on Facebook. Google plans to "fake generators" access to the distribution of advertising revenues (AdSense), provide search results with the mark "FactCheck" (with fact checking). There are ideas for attracting crowdfunding sites, blockchain technologies to combat fakes.

How to recognize fake? Fakes are often framed too well, too extreme. The following "recognizers" of fake news are distinguished, which can be implemented simply.

1. Note the domain and URL: trusted sites have familiar names and standard extensions, such as .com or .edu.
2. Read the section "About Us": its structure and "excessive" vocabulary may indicate an unreliable source.
3. Take a look at the quotes: good stories quote several experts to get different points of view on the problem.
4. Look who said: can you check the correctness of the quotes, is the source authoritative?
5. Check the comments: in social networks they can warn you when the story does not match the title.
6. Reverse image search: if an image used in a news story appears on other sites (on other topics), this is a sign that it does not match the fake.

Social networks are fake distributors. Platforms such as Facebook and Twitter allow you to easily share current news without wasting time critically evaluating it. Readers themselves are not inclined to critically evaluate the news, spreading fake on social networks with great speed (Persily, 2017).

### **Methodology**

The work uses methods of system analysis and synthesis, critical comparative analysis, mathematical modeling and others.

Methodologies, strategies to combat fakes are subject to the following principles.

1. Remember the problem: popular news aggregators and social networks compete for your attention, sometimes they can manipulate the viewer.
2. We think critically: we critically evaluate the news if it is too good (bad) to be true, probably it is. Most fakes are based on our desire to confirm our beliefs, positive or negative.
3. We compare facts with reliable sources. Find time to evaluate using authoritative sources, including media libraries. Although authoritative news sources themselves may sometimes be mistaken, they carefully check the facts to confirm their news.
4. We stop the spread of fakes: you can contribute to the non-proliferation of fakes further in social networks, by mail, etc.

The fight against fake news in social networks boils down to understanding the goals and platform of social networks. They make money, so often run ads tailored to our interests, search history. This is targeted advertising, including neuromarketing processes, methods.

It's important to know those who have business accounts in social networks. Knowing that news first passes through the filter of previously collected data, we can be more responsible. If you represent a social network business, a marketing platform, it's important that your messages correspond to the brand, create positive relations with customers.

Fake news on social networks seems inevitable. The best way to fight is to maintain a healthy curiosity for what is read in the tape, understand how platforms oversee content, and be "suspicious". Social networks are a powerful tool for business, including "private owners", if used consciously and competently.

Methods for identifying fake news are grouped by both news context content and social context content (Shu et al., 2021; Stella, Ferrara & Domenico, 2018). Fake news of the first class distinguish the quality of vocabulary (inconsistency, non-uniqueness, etc.), trust rating, age of the domain, etc. We identify the second class in context.

Figure 1 shows the distribution of television news consumption in comparison with computer news consumption (age 18-55). The y-axis is the average daily news consumption on the computer, the x-axis is the average news consumption on the desktop per day (Feldman, 2007). Shows a monthly picture of groups corresponding to different web news consumption ranges. The average consumption of TV news and the size of the group (percentages of all participants in the discussion) are estimated.

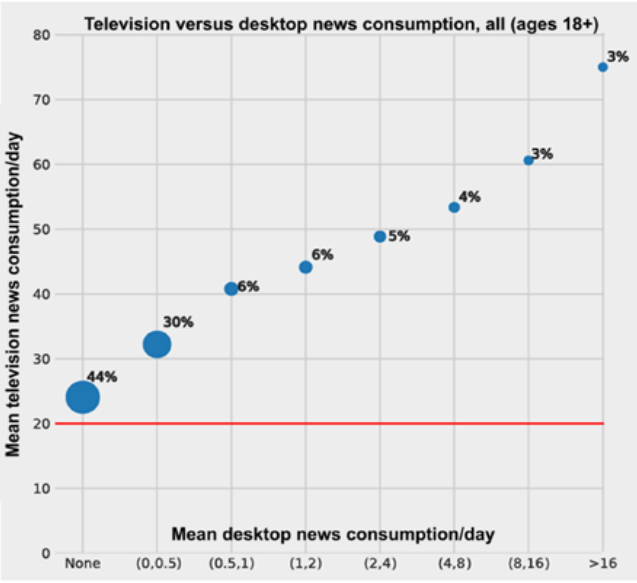


Fig. 1. Picture of television news consumption versus desktop news consumption by random samples (Feldman, 2007)

**Results**

Many fakes that are dangerous from the point of view of public influence are transmitted from one visitor to another by various mechanisms. If you generalize, distract from secondary parts, consider systemically, then you can limit yourself to two main mechanisms - horizontal and vertical transmission. With horizontal exposure to susceptible, it is carried out by intergroup (direct or indirect) contact with "contagious" fake accounts. Vertical impact - direct, intra-group.

It's assumed that all horizontal transmissions occur among adults, there is a T period during which new users do not participate in the horizontal transfer of fake (waiting period).

We need functions that specify the number of distributors, the number of susceptibilities at time t and evaluated at time t-T. We assume that spreading and sensitive are removed at a constant rate R and r, respectively, and adolescents with a frequency G and g, respectively. The model does not allow remote distributors to participate in the reproduction or transmission of fakes. We assume that adults susceptible have a constant birth rate b, and distributors have a birth rate with a coefficient B.

The vertical mechanism is introduced into the model by assuming that their proportion is equal to p (susceptible) and q=1-p (distributors), respectively. The growth of susceptibilities is given by

$$bS(t - T) + pBI(t - T),$$

and distributors - as

$$qBI(t - T),$$

where

$$b = \tilde{b} \exp(-gT),$$

$$B = \tilde{B} \exp(-GT).$$

Dynamic equations of the model - with delays:

$$\frac{dS(t)}{dt} = -rS(t) + bS(t - T) + pBI(t - T) - kS(t)I(t) \quad (t > 0),$$

$$dI(t)/dt = -RI(t) + qBI(t - T) + kS(t)I(t)$$

to which the initial conditions of the view must be attached

$$S(t) = S_0(t), \quad I(t) = I_0(t), \quad -T \leq t \leq 0,$$

The above model can be considered as a model in which S(t), I(t), respectively, the number of susceptible and distributors.

The model retained the possibility that r is not equal to R and b is not equal to B. It can be assumed that r=R. Fake at first usually does not reduce its performance, but suppose (it's natural) that after several fake cycles this happens. Therefore, it's interesting to consider both cases: b=B, b>B.

The only nonlinearity in the model is related to the transmission time kS(t)I(t). For one cycle, you can adjust, parametrically adapt the model. In the special case, at T=0, the model is reduced to the Cauchy problem for ordinary differential equations:

$$\frac{dS(t)}{dt} = (b - r)S + pBI - kSI,$$

$$\frac{dI(t)}{dt} = (qB - R)I + kSI.$$

The results on the qualitative behavior of the system described by the model are important. Consider isolated equilibrium solutions:

$$(0;0), \quad (S^*; I^*) = \left( \frac{R-qB}{k}, \frac{(R-qB)(b-r)}{k(R-B)} \right)$$

when R is not equal to B.

Solution (S\*; I\*) is of interest only when

$$R - qB \geq 0, \quad (b - r) / (R - B) \geq 0.$$

Similarly (Busenberg, Cooke, & Pozio, 1983) we break the results into 5 cases at p>0, k>0 and, depending on the values of the parameters, we formulate statements.

Theorem 1.

Case 1. If  $r < b, R > B$ , then (S\*; I\*) is feasible and globally stable with respect to all solutions with initial condition  $I(0) > 0$ .

Case 2. If  $r \geq b, R \geq B$  and one of them is not equal to the other, then  $I(t) \rightarrow 0$ , at  $t \rightarrow \infty$ ,  $S(t) \rightarrow 0$ , if  $b < r$  and positive.

Case 3. If  $r = b$ ,  $R = B$  and  $P(0) = S(0) + I(0)$ , then  $P(t) = P(0)$  for all  $t$  and  $I(t) \rightarrow 0$ , if  $pB/k > P(0)$  or  $I(0) = 0$ , since  $I(t) \rightarrow P(0) - pB/k$ , since  $pB/k \leq P(0)$  and  $I(0) > 0$ .

Case 4. If  $r < b, R \leq B$  or  $r \geq b, R \leq qB < B$ , then  $S(t) \rightarrow pB/k$  and  $I(t) \rightarrow \infty$  (except  $S = I = 0$ ).

Case 5. If  $r > b, qB < R < B$ , then  $(S^*; I^*)$  is a saddle point and there is a separatrix. For initial conditions below the separatrix,  $S(t) \rightarrow 0, I(t) \rightarrow 0$ , then for conditions above the separatrix we have  $I(t) \rightarrow \infty, I(t)/P(t) \rightarrow 1, t \rightarrow \infty$ .

### Discussion

It follows from the theorem that:

No periodic decisions;

- 1) there are two cases where the population is regulated to a constant size - case 1 (removal of susceptible below their activation, the opposite is true for infection) and there is a stable level of fake infection, which occurs regardless of the initial conditions  $S(0)$ ,  $I(0)$  and  $k, q$ , and there is no threshold or minimum community size necessary for the disease to become "endemic" (characteristic in the community);
- 2) in case 3, the total number in the community remains constant, there is a threshold condition for maintaining fakes;
- 3) in case 4, the community "explodes", as in case 5, when the thresholds  $S(0)$ ,  $I(0)$  are exceeded;
- 4) in case 5, when the rate of vertical propagation of fakes  $q$  increases, separatrix drops, it becomes more likely that the threshold is exceeded.

In the first approximation, we can take  $b = r, B = R$  (case 3 is applicable). Threshold condition

$$P(0) \geq (1 - q)B/k,$$

where  $q$  is vertical transmission rate and prevalence

$$1 - (1 - q)B/kP(0).$$

On the other hand, when averaging over several cycles, the  $R$  value can exceed the  $r$  value.

Cases 1-5 are possible with  $R > r$  and even with  $b=B$ , except case 5. For example, in case 1, if  $r < b = B < R$ , we see that process persists.

The system has a single solution, like a sequence of differential equations with certain initial conditions.

### Conclusion

People often tend to uncritically evaluate the news that their friends share (albeit virtual, from a community, group) or confirming their beliefs (albeit incorrect). Fakes contributed to important political, economic and social events of recent times. Especially the accompanying pandemic COVID-19.

In Russia, systematic research is not carried out on fake news in social networks, the blogosphere. The study is useful for analyzing and predicting, identifying and neutralizing fake news generators in social networks.

Episodes of fake infection can be described by more complex mathematical models and mechanisms. It is necessary to strengthen modeling, forecasting fake infection in the social network, media environment in order to counter it.

In the model considered, the prevalence of  $r$  has very simple behavior, it is necessary to investigate more complex models that will help to develop relevant measures to prevent the spread of fakes and neutralize the damage from this. The media literacy of the readership should also be relevant to this.



- Allen, J., Howland, B., Mobius, M., Rothschild, D., & Watts, D. (2020) Evaluating the fake news problem at the scale of the information ecosystem. *Science Advances*, 6(14), is. 3539. DOI: 10.1126/sciadv.aay3539
- Aral, S., & Eckles, D. (2019) Protecting elections from social media manipulation. *Science*, No. 365, pp. 858–861.
- Busenberg, S.N., Cooke, K.L., & Pozio, M.A. (1983) Analysis of a model of a vertically transmitted disease. *Journal of Mathematical Biology*, 17(3), pp. 305-329.
- Domagoj, B., & Volarević, M. (2018) New Problems, Old Solutions? A Critical Look on the Report of the High Level Expert Group on Fake News and On-Line Disinformation. *Media Studies*, № 9(17), pp. 106-117. DOI: 10.22363/2313-1438-2018-20-3-447-460.
- Feldman, L. (2007) The news about comedy: Young audiences, The Daily Show, and evolving notions of journalism. *Journalism*, No. 8, pp. 406-427.
- Golovatskaya, O. (2019) The meaning and origin of the term "Fake news". *Communicology*, 7(2), pp. 139-152. DOI 10.21453/2311-3065-2019-7-2-139-152.
- Kaziev, V., Kaziev, K., & Kazieva, B. (2017) Fundamentals of legal informatics and informatization of legal systems: a textbook. 2nd ed. M.: University textbook: INFRA-M, 336p.
- Lazer, D., Baum, M., Benkler, Y., Berinsky, A., Greenhill, K., et al. (2018). The science of fake news. *Science*, No. 359, pp. 1094-1096.
- Persily, N. (2017) Can democracy survive the Internet? *J. Democr.*, No. 28, pp. 63-76.
- Sukhodolov, A., & Bychkova, A. (2017) "Fake News" as a phenomenon of modern media space: concept, types, purpose, countermeasures. *Questions of the Theory and Practice of Journalism*, 6(2), pp. 143-169. DOI 10.17150/2308-6203.2017.6(2).143-169.
- Shu, K., Sliva, A., Wang, S., Tang, J., & Liu, H. (2021) Fake News Detection on Social Media: A Data Mining Perspective. *Arxiv.org*. URL: <https://arxiv.org/pdf/1708.01967.pdf> (Date of the application: 15.10.2020).
- Stella, M., Ferrara, E., & Domenico, M.D. (2018) Bots increase exposure to negative and inflammatory content in online social systems. *Proc. Natl. Acad. Sci. USA*, No. 115, pp. 12435-12440.
- Tretyakov, A., Filatova, O., Zhuk, D., Gorlushkina, N., & Puchkovskaya, A. (2018) Method of determining Russian-language fake news using elements of artificial intelligence. *Intern. J. of Open Information Technol.*, 6(12), pp. 99-105.
- Vosoughi, S., Roy, D., & Aral, S. (2018) The spread of true and false news online. *Science*, No. 359, pp. 1146–1151.