

Physical Models for Solar Cycle Predictions

Prantika Bhowmik¹ (10 · Jie Jiang^{2,3} (10 · Lisa Upton⁴ (10 · Alexandre Lemerle^{5,6} (10 · Dibyendu Nandy^{7,8} (10)

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Abstract

The dynamic activity of stars such as the Sun influences (exo)planetary space environments through modulation of stellar radiation, plasma wind, particle and magnetic fluxes. Energetic solar-stellar phenomena such as flares and coronal mass ejections act as transient perturbations giving rise to hazardous space weather. Magnetic fields - the primary driver of solar-stellar activity – are created via a magnetohydrodynamic dynamo mechanism within stellar convection zones. The dynamo mechanism in our host star – the Sun – is manifest in the cyclic appearance of magnetized sunspots on the solar surface. While sunspots have been directly observed for over four centuries, and theories of the origin of solar-stellar magnetism have been explored for over half a century, the inability to converge on the exact mechanism(s) governing cycle to cycle fluctuations and inconsistent predictions for the strength of future sunspot cycles have been challenging for models of the solar cycles. This review discusses observational constraints on the solar magnetic cycle with a focus on those relevant for cycle forecasting, elucidates recent physical insights which aid in understanding solar cycle variability, and presents advances in solar cycle predictions achieved via data-driven, physics-based models. The most successful prediction approaches support the Babcock-Leighton solar dynamo mechanism as the primary driver of solar cycle variability and reinforce the flux transport paradigm as a useful tool for modelling solar-stellar magnetism.

Keywords Solar magnetic fields \cdot Sunspots \cdot Solar dynamo \cdot Solar cycle predictions \cdot Magnetohydrodynamics

1 Introduction

The Sun's magnetic field is the primary determinant of the electromagnetic and particulate environment around our planet as well as the heliosphere. Solar magnetic field variability is manifested through different spatial and temporal scales: from long-term decadal-scale variations in open magnetic flux, 10.7 cm radio flux, and total solar irradiance to short-term sporadic energetic events such as flares and coronal mass ejections (CMEs). While longer-term variations depend on the distribution and evolution of the large-scale global magnetic

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field of the Sun, short-term perturbations like CMEs or flares originate primarily from localised magnetic structures of relatively smaller spatial scales. Despite the differences, solar magnetic variability of different spatial and temporal scales can intermingle and influence each other since all follow the same laws of physics. Ultimately, solar magnetic fields therefore are responsible for shaping space weather and space climate (Nandy et al. 2023).

High energy radiation and particle fluxes originating from extreme space weather events (flares and CMEs) can damage satellites orbiting the Earth and are hazardous to astronaut health. The impact of such events can harm critical infrastructures on the ground, resulting in direct or cascading failures across vital services such as communications and navigational networks, electric power grids, water supply, healthcare, transportation services etc. (Schrijver et al. 2015).

Flares and CMEs are linked to the complex magnetic field distribution on the solar surface, which is dictated by the emergence of sunspots and the subsequent evolution of the active region associated magnetic flux. Thus the frequency of short-lived energetic events depends on the number of sunspots emerging within a solar cycle. Simultaneously, the slower and longer-term evolution of the magnetic field of sunspots determines the amplitude of open magnetic flux and the speed and structure of solar wind emanating from the Sun, in effect, defining the space climate. It has direct consequences on the forcing of planetary atmospheres, the life span of orbiting satellites and planning of future space missions. Thus understanding and predicting phenomena which governs space weather and space climate is a scientific pursuit with immense societal relevance – in which solar cycle predictions occupy a central role (NRC 2013; Schrijver et al. 2015; 2017; NSWSAP 2019, 2022).

The methodologies for predicting different aspects of space weather and space climate are diverse but broadly relies upon observations, empirical methods, computational models and consideration of the physics of the system. Here we focus primarily on the last theme, i.e., developing our understanding of solar variability using the laws of physics towards attaining the goal of solar cycle predictions. Now physics-based prediction on different time scales itself is an extensive topic, and a complete narrative is beyond the scope of this review. Instead, we limit ourselves to decadal-centennial scale variability associated with the sunspot cycle. We emphasize that physical understanding gleaned from successful solar cycle prediction models also apply to other Sun-like stars with similar dynamo mechanisms.

Sunspots are understood to be the product of a dynamo mechanism operating within the Sun's convection zone (SCZ, hereafter) dictated by the laws of magnetohydrodynamics (Charbonneau 2020). In the SCZ, the kinetic energy stored in the ionised plasma converts to the magnetic energy primarily stored in the toroidal component of the magnetic field. The toroidal field, following significant amplification, rises through the SCZ due to magnetic buoyancy (Parker 1955) and emerges on the solar surface as strong localised magnetic field concentrations, forming Bipolar Magnetic Regions (BMRs, primarily), of which the larger ones are optically identified as sunspots. One of the mechanisms that contribute to poloidal field generation is the mean-field α -effect which relies on helical turbulent convection twisting rising toroidal fields whose net impact is to produce a finite poloidal component. On the surface, observed small and large-scale plasma motion redistributes the magnetic field associated with the BMRs, resulting in the reversal and growth of the existing global magnetic dipole moment (poloidal component) of the Sun. This process – termed as the Babcock-Leighton (B-L, hereafter) mechanism (Babcock 1961; Leighton 1969) – is another means to generate the poloidal component.

The strength of the magnetic dipole at the end of a solar cycle is found to be one of the best precursors for predicting the amplitude of the following cycle. This in itself is related to the stretching of the poloidal field through the deterministic process of differential rotation.



However, observations, analytic theory and data-driven models of decadal-centennial scale variability in the solar dynamo mechanism indicates that the B-L mechanism is the primary source of variability in the poloidal field and hence in the sunspot cycle (Cameron and Schüssler 2015; Bhowmik and Nandy 2018).

Any physics-based model aiming for solar cycle predictions must be dictated by the laws of magnetohydrodynamics, contain the essence of the dynamo mechanism, and be constrained by observed plasma and magnetic field properties in the solar surface and the interior. Some recent studies (Petrovay 2020; Nandy 2021) have explored the diversity of methods employed for sunspot cycle predictions and their credibility. Compared to these studies, this review will primarily focus on physics-based predictions of the sunspot cycle. Also see Jiang et al. (2023) for a comparison of cycle 25 predictions using physics-based models.

In the following sections, we begin with a brief account of the observed distribution of magnetic field (primarily on the solar surface) and plasma flows which serve as building blocks and constraints for computational models (Sect. 2). This is followed by a short description of the computational models of magnetic field evolution on the solar surface and interior which has shown great promise as predictive models of the solar cycle (Sect. 3). Physical insights on sources of irregularities in the strength of the solar cycle and amplitude modulation mechanisms – which are gleaned from simulations and attempts to match observations – are discussed in Sect. 4. In Sect. 5 we present a review of physics-based solar cycle predictions limiting ourselves to data-driven modelling approaches. We conclude in Sect. 6 with a discussion on the relevance of solar cycle predictions models for theories of solar-stellar magnetism and end with some remarks on future prospects in the field of solar cycle predictability.

2 Constraints from Solar Observation

Solar magnetic field observations are primarily limited to the visible layer of the Sun, i.e., the photosphere. Plasma flows are observed both on the surface as well as in the interior through inference using tools of helioseismology. Computational models utilize these information for purposes of constraining and calibrating the models. The main goals driving this synergy between observations and models are to achieve a better understanding of the physical processes ongoing in the Sun and develop predictive models of solar activity. In this section, we focus on observations which are relevant to surface flux transport (SFT) and dynamo models of the solar magnetic field. For a detailed account of solar observations, see Norton et al. (2023), this collection.

2.1 Sunspot Number

Although sunspots have been observed systematically through telescopes from the early 1600's, in the early 1800's solar astronomer Samuel Heinrich Schwabe began plotting the number of sunspots as a function of time and discovered that their appearance was cyclic with a period of about eleven years (Schwabe 1844). As new phenomena (e.g., flares, CMEs, etc.) on the Sun were discovered, it was found that they too varied along with the sunspot number. Solar activity is often characterized by the Monthly Sunspot Number, a count of the number of Sunspots or Active Regions observed each month as a function of time. The official count is maintained by the Sunspot Index and Long-term Solar Observations (SILSO) at



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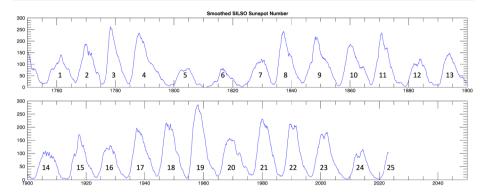


Fig. 1 The SILSO Sunspot Number. The smoothed SILSO Monthly mean total sunspot number (v2.0), smoothed illustrates the rise and fall of solar activity from 1750 to the present (marked with solar cycle numbers)

the Royal Observatory of Belgium, Brussels. In 2015, the sunspot number data were revised to version 2.0, to account for changes in observers and provide a more consistent data series throughout the historical record (Clette et al. 2014). Version 2.0 of the SILSO Monthly mean total sunspot number, smoothed with a 13 month Gaussian filter, is illustrated in Fig. 1.

The SILSO data series now shows nearly 25 solar cycles. Each solar cycle has a period of about 11 years and an average amplitude (v2.0) of 180, with a range of about 80 (e.g., Solar Cycle 5) to 280 (e.g., Solar Cycle 19). The length of the cycle correlates with the amplitude of the cycle such that bigger cycles tend to be shorter in duration and weaker cycles tend to be longer (Waldmeier 1935). The shape of the solar cycle approximately appears as an asymmetric Gaussian function, with a rapid rising phase and a longer decaying phase. However, for a better fitting with the observed sunspot cycle, combinations of power law and exponential functions have also been utilized (Hathaway et al. 1994; Jiang et al. 2018). Shorter term variability in the cycle, on the order of about 2 years, causes many cycles to have two or more peaks, which are often more pronounced in weaker cycles (Karak et al. 2018). Sunspot cycles have other irregularities too, for example, the Sun entered into a prolonged near-minimum state during 1645 – 1715 (this was pointed out by G. Spörer and E. W. Maunder in the 1890s). This phase, known as the Maunder Minimum (Eddy 1976), is a period where the solar activity cycle was operating at an exceptionally weak state for several decades.

2.2 Magnetic Field Observations Relevant for Solar Cycle Models

Perhaps one of the most significant solar discoveries in the twentieth century, was the realization that sunspots are magnetic in nature (Hale 1908b,a). Sunspots are now known to host very strong magnetic fields on the order of 1000 G and often appear as a collection of spots. The magnetic counterparts of sunspots correspond to Active Regions. An Active Region, in general, appears as a pair of spots (with opposite polarities) which are referred to as BMR. Sunspots are the optical counterparts of active regions. Besides the spots, the background surface field has a strength of only a few Gauss. The dynamics of the Sun's surface magnetic field is captured by the 'Magnetic Butterfly Diagram', see Fig. 2. This figure is created by plotting the longitudinally average radial magnetic field on the Sun as a function of latitude

¹https://www.sidc.be/silso/home.



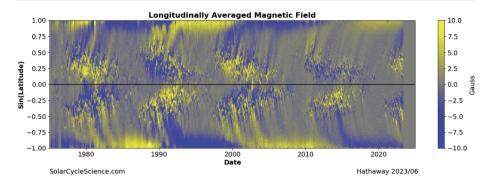


Fig. 2 The magnetic butterfly diagram illustrates the evolution of magnetic flux on the Sun over several cycles, providing observational constraints that govern models used to make solar cycle predictions. The SOLIS/MDI/HMI magnetic field data averaged over each Carrington Rotation is plotted as a function of latitude and time, with the positive (negative) polarity shown in yellow (blue). Figure courtesy D.H. Hathaway via www.solarcyclescience.com

and time. This figure illustrates several properties that serve as constraints for the SFT and dynamo models and which are tests for predictive models. The properties include:

- Sporer's Law: Active Regions begin emerging at mid-latitudes (~ 30 degrees). As the
 cycle progresses, active regions form bands of flux that moves equator-ward (Carrington
 1858).
- Joy's Law for Tilt Angles: Active Regions tend to have a characteristic tilt such that the angle between the local parallel of latitude and the line joining the leading polarity spot (which appears closer to the equator) and the following polarity spot statistically increases with increasing latitude (Hale et al. 1919).
- Hale's Polarity Law: The relative polarity of Active Regions (between the leading and following spots) is opposite across the equator, and this polarity changes sign from one cycle to the next (Hale et al. 1919). Thus a complete magnetic cycle has a periodicity of 22 years.
- Polar Fields: In addition to the flux in emerging Active Regions, the Sun possesses unipolar concentrations of magnetic flux near both of the poles. The polarity is opposite across hemispheres, reverses polarity at about the time of solar cycle maximum (Babcock and Livingston 1958) and attains its maximum amplitude around cycle minimum. This large-scale field is known as the 'Polar field', and its evolution plays an important role in solar cycle predictions. However, due to projection effects, polar field measurements suffer from erroneous values, thus prompting the need for off-ecliptic space missions focusing on the Sun's poles (Nandy et al. 2022).
- Magnetic Field Surges: Streams of weak polarity flux are carried from the active latitude to the poles in about 2–3 years. These streams are responsible for the build-up and reversal of the polar fields (Babcock 1961). The strength of these surges reaching the poles can vary significantly based on the associated active regions. The emergence latitudes, tilt and flux of the active region, and frequency of active region emergence are all important factors in determining how much a given active region will contribute to the polar field evolution. The high latitudes emergence with large tilt (Yeates et al. 2015) and long-lasting activity complexes (Wang et al. 2020) tend to generate prominent poleward surges. However, the surges resulting from high latitude emergence usually only bring the transient influence of polar field (Jiang et al. 2014a; Yeates et al. 2015). On the contrary, 'rogue region' (e.g., with a large area and high tilt angle) emerging at low latitudes



usually cannot generate prominent surges but can influence the polar field development significantly (Nagy et al. 2017).

The redistribution of the active regions' magnetic flux across the solar surface and the interior convection zone happens through the collective effect of small and large-scale plasma motions which provide additional constraints on models.

2.3 Plasma Flows

Plasma flows in the solar convection zone may be divided into three categories based on the physical role they play in the solar dynamo mechanism: convective flows, differential rotation, and meridional circulation. The thermal flux through the solar convection zone and consequent temperature gradient causes the plasma within the solar convection zone to rise to upper layers, transfer or radiate their energy away and sink back down after cooling. As a result, convective cells with a spectrum of different scales (Hathaway et al. 2015) are formed ranging from granules (diameter ~ 1 Mm) with lifetimes of minutes to hours, to supergranules (diameter ~ 30 Mm) with lifetimes of days, and to the largest convective structures (diameter ~ 200 Mm) with lifetimes of months. These convective motions are turbulent in nature and effectively distribute the magnetic field in the entire convection zone, including the solar surface, similar to a diffusive process.

The Sun rotates differentially which was first found by tracking sunspots on the solar surface (Adams 1911; Belopolsky 1933; Howard 1984). This differential rotation at the surface is such that the large-scale plasma flow speed along the direction of solar rotation varies latitudinally with a faster-rotating equator than the poles. Later, helioseismology (Schou et al. 1998; Basu 2016) was utilized to obtain the structure and spatial variation of rotation rate inside the solar convection zone. The radiative zone rotates as a solid body resulting in a strong radial shear within the tachocline which is thought to encompass the (stable) overshoot layer at the base of the convection zone. The differential rotation plays a crucial role in the generation and amplification of the toroidal component of the Sun's large-scale magnetic field (see Sect. 3).

Another large-scale subsurface plasma flow known as the meridional circulation (Hanasoge 2022) which in near-surface layers carries plasma from the equatorial region to the poles (in both hemispheres) with a varying speed dependent on latitude. The flow speed becomes zero at the equator and the poles, and the circulation attains its peak speed $(10-20 \text{ m s}^{-1})$, about 1% of the mean solar rotation rate) near mid-latitude. The law of mass conservation dictates an equator-ward return flow of plasma deeper inside the solar convection zone, which, however, remained hard to map using helioseismic observations due to its small amplitude. While some recent studies (Rajaguru and Antia 2015; Liang et al. 2018) have suggested that meridional circulation is a single-cell flow where the return flow is at the depth of the solar convection zone (depth $< 0.8 R_{\odot}$), others (Hathaway 2012; Zhao et al. 2013; Hathaway et al. 2022) suggest that it may be multi-cellular in depth and or latitude. The shape of the meridional profile in latitude and radius is crucial in determining the various properties of the sunspot cycles, including cycle duration. Early flux transport dynamo models suggest that a deep single-cell meridional flow threading the convection zone is necessary to match solar cycle observations (Nandy and Choudhuri 2002; Hazra et al. 2014). However, the recent Babcock-Leighton-type dynamo models (Zhang and Jiang 2022; Zhang et al. 2022) indicate that the deep single-cell meridional flow is not crucial to match solar cycle observations.

Note that both small-scale and large-scale plasma flows are not static. Helioseismic observation shows that the Sun's differential rotation varies with time in an oscillatory fashion



and is correlated with the solar activity cycle – it is known as the solar torsional oscillation (Zhao and Kosovichev 2004; Howe 2009). Meridional circulation also exhibits cycle-dependent temporal variation in its peak speed, with reduced amplitude during cycle maximum compared to the minimum (Hathaway and Rightmire 2010; Hathaway and Upton 2014; Hathaway et al. 2022). However, for both large-scale plasma flows, such variations constitute less than 20% of the average profiles. Thus, computational models with time-independent plasma flow profiles can reproduce the majority of the observed large-scale magnetic field variability.

2.4 Polar Fields as Precursors of the Strength of Sunspot Cycles

The temporal evolution of the averaged polar field has a $\pi/2$ phase difference with the sunspot cycle. As mentioned earlier, the average polar field strength at cycle minimum serves as an important element in predicting sunspot cycles. Although direct observation of the polar fields became available only in the 1970s, indirect measures of polar flux exist based on proxies. Polar flux evolution derived from polar faculae observations (Muñoz-Jaramillo et al. 2012) cover a period of 100 years (during cycles 14-24). Note that the average polar flux during cycle minimum is a close representation of the Sun's magnetic dipole (axial) moment – which acts as a seed to generate the following solar cycle. In fact, the average polar flux at the nth cycle minimum has the maximum positive correlation with the amplitude of the n+1th cycle [see, Fig. 3]. The correlation decreases drastically for the amplitude of cycles nth, n+2th and n+3th as depicted in Fig. 3. Figure 3 reflects on two crucial aspects of solar cycle predictability: first, a strong solar cycle does not result in a strong polar field generation at that cycle minimum [see, Fig. 3(a)] and the memory of the polar field (of nth cycle) diminishes beyond the next $(n+1^{th})$ cycle [see, Fig. 3(c) and (d)]. It is important to note here that these empirical observational evidences especially the decreasing correlation beyond the immediate cycle, were preceded by flux transport dynamo models exploring the memory issue which predicted that the sunspot cycle memory is limited primarily to one cycle alone in certain parameter regimes related to flux transport timescales (Yeates et al. 2008; Karak and Nandy 2012).

3 Physical Modeling Approaches

In an astrophysical magnetised plasma system like our Sun, we expect the following properties of the plasma will be satisfied: the velocity is non-relativistic, the collisional mean free path of the atomic or molecular constituents of the plasma is much shorter than competing plasma length scales, and the plasma is electrically neutral and non-degenerate. In such a system, the evolution of the magnetic field is dictated by the magnetohydrodynamic (MHD) induction equation, which is a combination of Ohm's law, Ampère's law and Maxwell's equations:

$$\frac{\partial \mathbf{B}}{\partial t} = \nabla \times (\mathbf{u} \times \mathbf{B} - \eta \nabla \times \mathbf{B}). \tag{1}$$

Here ${\bf u}$ and ${\bf B}$ are the plasma velocity and magnetic fields, respectively, and $\eta=1/\mu\sigma$ is the magnetic diffusivity, with μ the magnetic permeability and σ the electric conductivity. Additionally, the magnetic field satisfies the divergence-free condition, $\nabla \cdot {\bf B}=0$. The spatio-temporal evolution of the plasma flow is dictated by Navier–Stokes equation,

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla)\mathbf{u} = -\frac{1}{\rho}\nabla P + \mathbf{g} + \frac{1}{\rho}(\mathbf{J} \times \mathbf{B}) + \nu \nabla^2 \mathbf{u}, \tag{2}$$



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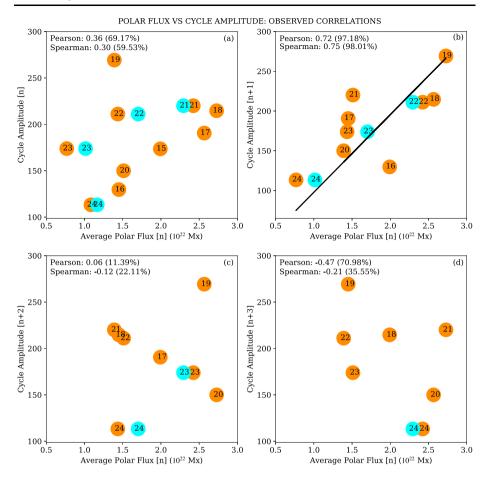


Fig. 3 Observed cycle-to-cycle correlations between the polar flux at cycle minima (say, [n]) and the cycle amplitude of different cycles, namely (a) cycle [n], (b) cycle [n+1], (c) cycle [n+2], and (d) cycle [n+3]. The numbers inside the circles indicate the associated solar cycle numbers. The colors of the circles differ based on the source of polar flux data, orange: averaged polar flux obtained from polar faculae count, cyan: the average dipole moment (scaled appropriately to place them in the figure). Image reproduced with permission from Nandy (2021) copyright by Springer Link

where ρ is plasma density, P is plasma pressure, g is the gravitational acceleration, $\mathbf{J} = (\nabla \times \mathbf{B})/\mu$ is the electric current, and ν is kinematic viscosity. Additionally, the plasma flow obeys mass conservation through the continuity equation. Along with equations (1) and (2), one should take into account the conservation of energy and equation of states involving the pressure and plasma density. In an isolated system, where we can ignore the Poynting flux, the mechanical energy stored in the flow (u) acting in the opposite direction of the Lorentz force ($\mathbf{J} \times \mathbf{B}$) is converted to magnetic energy. This magnetic energy can decay through the dissipation of the electrical currents supporting the magnetic field.

Thus the sustainability of a dynamo depends on the relative strength between the induction effect controlled by the velocity field (the first term on the R.H.S in equation (1) and the Ohmic dissipation (the second term on the R.H.S in equation (1). The ratio of these two terms is known as the magnetic Reynolds number, $R_m = uL/\eta$, where L is the length scale



determining whether inductive effect overcomes the dissipative processes. In most astrophysical systems, a very large L ensures a very high $R_{\rm m}$, which is crucial for the survival and growth of the dynamo.

In an ideal scenario, solving the complete set of MHD equations associated with the conservation of mass, momentum, energy, and magnetic flux including the magnetic induction equation in the SCZ should provide the Sun-like spatio-temporal evolution of the velocity and magnetic field with the given Sun-like plasma properties. However, this requires the numerical models to be capable of comprising a wide range of spatial and temporal scales characterizing fluid turbulence at high viscous and magnetic Reynolds number medium — which is quite challenging from the computational point of view. While with increasing computational power and improved algorithms, full MHD models are becoming more realistic, the parameter regimes are still nowhere near the real solar convection zone. Moreover, all the existing MHD models operate with enhanced dissipation, much stronger than the characteristic dissipation in the solar interior. A comprehensive account of the MHD simulations of solar dynamos is presented in Käpylä et al. (2023), this collection, thus, we restrain ourselves from going into further details.

The scope of the growth of the dynamo is encapsulated within the advective part of the induction equation $[\nabla \times (\mathbf{u} \times \mathbf{B})]$ in Eq. (1), where any pre-existing magnetic field (B) is amplified by the plasma flow through the shearing term $[\mathbf{B} \cdot \nabla(\mathbf{u})]$, compression and rarefication $[\mathbf{B}(\nabla \cdot \mathbf{u})]$, and advection $[(\mathbf{u} \cdot \nabla)\mathbf{B}]$. While any positive gradient in the plasma flow ensures growth of B, the dynamo-generated magnetic field should have the following observed characteristics (see Sect. 2) in the solar context:

- The large-scale magnetic field (the dipole component) should reverse on a decadal scale.
- The sunspot generating field component should have a $\pi/2$ phase difference with the dipole component, should exhibit an equator-ward migration, and the associated polarity should be anti-symmetric across the equator.
- On the solar surface, the dynamo model is expected to reproduce observed features of sunspots and the associated flux evolution, which include pole-ward migration of the diffused field and generation of observed polar field (as seen in Fig. 2).
- Moreover, the solar dynamo models should result in amplitude fluctuations in both the sunspot-generating component and the large-scale dipole component, along with observed empirical patterns and correlations between them.

Reproducing all these intricate details of the observed solar magnetic field and the velocity field while solving the full set of MHD equations in the turbulent convection zone indeed becomes a challenging problem. Thus one major and very successful alternative approach in the dynamo community has been to focus on the evolution of the magnetic field only by solving the induction equation (1) while utilizing prescribed plasma flow derived from observation (Charbonneau 2020). These are often termed as kinematic dynamo models. Another modelling approach, namely Surface Flux Transport (SFT) models, simulate only one half of the solar cycle, namely, the evolution of magnetic fields which have emerged on the solar surface mediated via prescribed flow parameters which are constrained by observations. We discuss them briefly below.

3.1 Solar Surface Flux Transport Models as Tools for Polar Field Predictions

The genesis of solar surface magnetic field evolution models, as well as the Babcock-Leighton mechanism for polar field generation can be traced to the heuristic ideas first proposed by Babcock (1961). Babcock attempted to explain the behavior of the large-scale



solar magnetic fields through a phenomenological description of the topology of the Sun's magnetic field and its evolution which was related to the emergence of systematically tilted BMRs and the subsequent diffusion of their flux, cross-equatorial cancellation and migration to the poles – culminating in the large-scale dipolar field reversal. This process was envisaged to be complemented by the stretching of the submerged large-scale flux systems by differential rotation to produce the sunspot forming toroidal field. Later, Leighton (1964) put these ideas on a firmer theoretical foundation. He suggested that the radial magnetic field at the surface of the Sun is advected and diffused kinematically like a passive scalar field.

The computational models capturing the evolution of this radial magnetic field $[B_r(\theta,\phi,R_\odot)]$ associated with the active regions are known as Surface Flux Transport (SFT) models. The temporal evolution of the longitudinal averaged radial field obtained from such simulations should have the distinct features observed in the magnetic butterfly diagram (Fig. 2). Then the SFT mechanism may also be derived from the MHD induction equation (1) as the time evolution of the radial component of the magnetic field B_r , evaluated at $r = R_\odot$, as:

$$\frac{\partial B_r}{\partial t} = -\frac{1}{r\sin\theta} \frac{\partial}{\partial\theta} \left(u_\theta B_r \sin\theta \right) - \frac{1}{r\sin\theta} \frac{\partial}{\partial\phi} \left(u_\phi B_r \right) + \eta_T \nabla^2 B_r. \tag{3}$$

Here, u_{θ} and u_{ϕ} denote two large-scale plasma flows on the solar surface: meridional circulation and differential rotation, respectively. The diffusivity η_T represents a simplification of the effect of turbulent convective motions of the plasma on B_r . Note that we haven't considered the contribution from radial diffusion, which appears in the mean-field MHD formulation of SFT (Yeates et al. 2023). To the linear formulation above 3, a source term must be added to account for the additional influx of magnetic field associated with the emergence of active regions, which are primarily BMRs. For a detailed description of SFT models and their theoretical considerations, refer to Jiang et al. (2014b), Yeates et al. (2023).

3.1.1 Genesis of Surface Flux Transport Simulations

Following the pioneering work by Babcock (1961) and Leighton (1964) describing the evolution of B_r on the solar surface, DeVore et al. (1984) created the first SFT model of the Sun. Their SFT model was originally used to constrain meridional flow at the surface, which was difficult to measure and very uncertain at that time. To mimic the emergence of active regions on the solar surface, Sheeley et al. (1985) included bipolar active region sources based on observed statistics. Wang et al. (1989) explored the role of surface flux transport and dissipation processes such as differential rotation, meridional flow, and diffusion to investigate their role in the reversal and build-up of the polar fields. They found that a) differential rotation was essential for separating the leading and following polarity flux in bipolar active regions, b) diffusion played a crucial role in cross-equatorial flux cancellation of the leading polarities and b) meridional flow was essential for transporting the following polarity flux to the poles aiding in polar field reversal and build-up. The primary physical ingredients of the surface processes resulting in the observed solar cycle associated polar field dynamics were now in place.

3.1.2 Evolution Towards Data Driven Surface Flux Transport Models

More evolved SFT models now have the ability to incorporate the observed flows (static and time-evolving) and assimilate data for realistic simulations of solar surface field evolution and polar field predictions. These models are also paving the way for realistic coronal field



and heliospheric modeling by providing time-dependent, data assimilated boundary conditions at the solar photosphere.

While many modern SFT models continue to parameterize the small-scale convective motions with a diffusivity coefficient, a novel class of magnetoconvective SFT (mSFT) models have been developed which emulate not only the spreading motions of convection outflows, but also the concentration of the magnetic network formed at the boundaries of convective structures. The first attempt at this was achieved by introducing a random attractor matrix (Worden and Harvey 2000) to replace the diffusivity. The attractor method was later adapted by the Air Force Data-Assimilative Photospheric Flux Transport model (ADAPT) (Arge et al. 2010; Hickmann et al. 2015). Another SFT model invoked a collision and fragmentation algorithm (Schrijver 2001). An alternative approach known as the Advective Flux Transport (AFT) (Upton and Hathaway 2014) model has been developed which mimics surface convection through spherical harmonics to generate an evolving velocity flow field that reproduces the size, velocities, and lifetimes of the observed convective spectrum (Hathaway et al. 2015).

Another major advancement in SFT models, brought about by the space-based Doppler-Magnetographs, is the availability of high cadence and high-resolution magnetograms. Schrijver and De Rosa (2003) were one of the firsts to directly assimilate magnetogram data into the SFT model by incorporating SOHO/MDI magnetic field observations within 60° of the disk center. Using their SFT maps as an inner boundary condition to a PFSS model, they were able to create an accurate reconstruction of the interplanetary magnetic field (IMF). More formal data assimilation processes (e.g., Kalman filtering) require that the observed data be merged with the simulated data in a way that accounts for the uncertainties. ADAPT (Hickmann et al. 2015) and AFT (Upton and Hathaway 2014) type SFT models employ Kalman filtering. The ADAPT model is used in conjunction with WSA-ENLIL model to aid in Space Weather Predictions. Since the surface field distribution drives the coronal magnetic field, SFT (and AFT) models have shown great capabilities for coronal field simulations and predictions (Nandy et al. 2018; Dash et al. 2020; Mikić et al. 2018; Mackay and Upton 2022; Yeates and Bhowmik 2022).

Studies are illuminating the influence that flow variations have on polar field dynamics. Hathaway and Upton (2014) found that variations in the meridional flow had a significant impact (\sim 20%) on the polar field strength. Cross equatorial flow (Komm 2022) significantly influenced the residual magnetic flux transported to the poles. These simulations clearly show that despite being relatively weak, the shape and amplitude of the meridional circulation is a crucial element shaping the solar cycle dynamics.

Incorporating the observed sunspots statistics (Jiang et al. 2011) on the solar surface is crucial – where the active region's emergence latitude and associated tilt angle and magnetic flux become major deciding factors to the final contribution to the dipole moment evolution (Petrovay et al. 2020a; Wang et al. 2021). It may appear that the final dipole moment at the end of a cycle (which acts as a seed to the following cycle) will then be deterministic – a strong cycle producing a strong poloidal field at cycle minimum. However, observation suggests otherwise [see Fig. 3(a)]: saturation in the final dipole moment, which is linked with two factors: tilt-quenching (Dasi-Espuig et al. 2010; Jiang et al. 2011) and latitude-quenching (Jiang 2020) (see Sect. 4.3.2 for more details). Moreover, scatter in the active region tilt angles (in addition to the systematic tilt according to Joy's law) introduces substantial uncertainty in the final dipole moment at cycle minimum (Jiang et al. 2014a; Bhowmik 2019). For example, a few big rogue active regions emerging at low latitudes with a "wrong" (i.e., opposite to the majority for this cycle) tilt angles can reduce the dipole moment amplitude, thus weakening the seed for the following cycle (Jiang et al. 2015; Nagy et al. 2017).



Observationally constrained and flux calibrated SFT models can now match decadal to centennial-scale solar surface magnetic field dynamics and are being used for predicting the polar field amplitude based on synthetic data inputs of the declining phase of the cycle – with a reasonably good degree of success (Cameron et al. 2010; Upton and Hathaway 2014; Cameron et al. 2016; Bhowmik and Nandy 2018; Wang et al. 2002). These models have become useful tools for understanding solar surface flux transport dynamics, exploring the nuances of the Babcock-Leighton mechanism for solar poloidal field generation, and are being coupled to data-driven dynamo models for predicting the strength of the sunspot cycle.

3.2 Flux Transport Dynamo Models as a Tool for Sunspot Cycle Predictions

Kinematic or flux transport dynamo models have shown exceptional fidelity for being used as tools for solar cycle predictions. The utility of these models are due to the possibility of prescribing observationally constrained, or theoretically "expected" velocity profiles (**u**) and assimilating observations of the poloidal field to obtain the spatio-temporal evolution of the solar magnetic field (**B**) by using Eq. (1). These models use two large-scale time-independent velocity profiles to incorporate the observed differential rotation and meridional circulation (see Sect. 2).

Based on the observed properties of the surface field, the large-scale solar magnetic field at cycle minimum can be reasonably approximated to be axisymmetric (independent of ϕ) and antisymmetric across the equatorial plane. This simplifies the kinematic dynamo problem further. Thus in spherical polar coordinates (r, θ, ϕ) , the magnetic field (\mathbf{B}) can be expressed as,

$$\mathbf{B}(r,\theta,t) = \nabla \times \mathcal{A}(r,\theta,t)\,\hat{\mathbf{e}}_{\phi} + \mathcal{B}(r,\theta,t)\,\hat{\mathbf{e}}_{\phi}.\tag{4}$$

The first term in the R.H.S. of the above equation is the poloidal component (\mathbf{B}_P , hereafter) in the meridional plane expressed through a vector potential (\mathcal{A}) and the second term (\mathcal{B}) corresponds to the toroidal component (\mathbf{B}_T , hereafter). The velocity can also be expressed similarly as a combination of the poloidal (meridional circulation) and toroidal (differential rotation) components. All these simplifications lead us to two separate but coupled equations for \mathbf{B}_P and \mathbf{B}_T , where the first corresponds to the axial dipole moment (or averaged polar field), and the latter is related to the sunspot-generating strong magnetic field.

The solution to the set of equations produces B_P and B_T with a $\pi/2$ phase difference, both having roughly decadal-scale periodicity (considering amplitude only). It is reflected in Fig. 4 showing the observed evolution of averaged polar field and sunspot cycles over four decades. A cycle begins with the strongest B_P and the weakest B_T . In the rising phase of a cycle, with increasing B_T (i.e., more sunspots of opposite polarity), B_P (i.e., average polar field) weakens gradually through the B-L mechanism. B_P changes its polarity during the cycle maximum and progresses towards a new maximum value (with opposite polarity) during the declining phase of the cycle while B_T continues to decrease till the cycle minimum. Polarity-wise, B_P and B_T have a 22-year-long periodicity, which is also evinced through Hale's polarity law (as discussed in Sect. 2). In the following section, we describe how the generation process of the two components of the Sun's magnetic field rely on each other.

3.2.1 Poloidal to Toroidal Field

The induction equation for the toroidal component $(\mathbf{B_T})$ includes a source term originating from the differential rotation $[\Omega(r, \theta)]$ in the SCZ, compared to which the sink term due



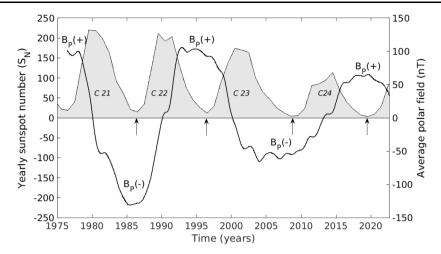


Fig. 4 The temporal evolution of yearly sunspot number (source: WDC-SILSO, Royal Observatory of Belgium, Brussels) and average polar field (source: Wilcox Solar Observatory) during sunspot cycles: 21-24. The arrows denote the epochs of cycle minima approximately. The sign of the average polar field corresponds to the sign of $B_P(+/-)$, and the sunspot number is related to the amplitude of B_T

to diffusive decay is negligible. Thus any pre-existing B_P will be amplified through shearing along the azimuthal direction (ϕ) by $B_P \cdot \nabla \Omega$ and generate new B_T . The differential rotation in the solar convection zone and the stable overshoot layer at its base (coinciding with the tachocline where turbulent diffusivity is suppressed) plays important roles in the amplification and storage of B_T (Muñoz-Jaramillo et al. 2009).

Following sufficient amplification by differential rotation, \mathbf{B}_T satisfies the magnetic buoyancy condition (Jouve et al. 2018; Fan 2021). Any perturbed part of the toroidal field rises as a flux rope through the SCZ, where it encounters the Coriolis force and turbulent diffusivity in the medium (Weber et al. 2011). The flux tube which eventually emerges through the solar surface creates a pair of spots (in general, following Hale's polarity rule) with a certain latitude-dependent tilt angle (following Joy's law), a somewhat fragmented structure and a reduced strength compared to its initial amplitude. A more detailed account of active region emergence has been discussed in Weber et al. (2023).

3.2.2 Toroidal to Poloidal Field Generation

Strictly axisymmetric flows and fields cannot sustain the dynamo process (Cowling's theorem). Thus to sustain a dynamo, a non-axisymmetric process must be invoked. Elaborate studies on kinematic dynamo models have utilized different mechanisms to convert \mathbf{B}_T to \mathbf{B}_P (Cameron et al. 2017; Charbonneau 2020) by utilizing intrinsically non-axisymmetric processes which are parameterized in the dynamo equations. We present below a very brief narrative of such approaches.

Turbulence and Mean-Field Electrodynamics Approach The thermally driven environment in the SCZ results in turbulent motion of the plasma, which therefore has a mean large-scale flow along with a fluctuating component $[\mathbf{u} = \langle \mathbf{u} \rangle + \mathbf{u}']$. While the mean component, $\langle \mathbf{u} \rangle$, corresponds to the standard axisymmetric large-scale plasma velocity (differential rotation and meridional circulation), the fluctuating component, \mathbf{u}' , vanishes when averaged



in the azimuthal direction. The magnetic field can be decomposed in a similar fashion: $\mathbf{B} = \langle \mathbf{B} \rangle + \mathbf{B}'$. Although the fluctuating parts of the velocity and the magnetic field vanish individually when averaged azimuthally, their product, $\mathcal{E} = \langle \mathbf{u}' \times \mathbf{B}' \rangle$ will sustain and overcome the restriction set by Cowling's theorem. \mathcal{E} is known as the mean turbulent electromotive force, and part of it serves as a source term (the α -effect) in the induction equation of \mathbf{B}_P . From a physical point of view, it can be linked to the helical twisting of the toroidal field component (\mathbf{B}_T) by helical turbulent convection. For a thorough description, please refer to Brandenburg et al. (2023), Hazra et al. (2023), this collection.

The Babcock-Leighton Mechanism The magnetic axis connecting the opposite polarities of active regions has a certain tilt with respect to the east-west (toroidal) direction which arises due to the action of the Coriolis force on buoyantly rising toroidal flux tubes (an inherently non-axisymmetric process, see Fisher et al. (2000) and references within). Thus, all active regions have non-zero components of magnetic moments along the north-south (poloidal) direction – which collectively contributes to the axial dipole moment generation and evolution (Petrovay et al. 2020a). Section 3.1 describes how the magnetic flux initially concentrated within tilted active regions decay and redistributes across the solar surface to generate the large-scale magnetic field. Thus is the so called Babcock-Leighton mechanism which converts $\bf B_T$ to $\bf B_P$. Observational evidence not only strongly supports the B-L mechanism, they also help constrain data-driven predictive SFT and dynamo models (Passos et al. 2014; Cameron and Schüssler 2015; Bhowmik and Nandy 2018; Bhowmik 2019). For a more detailed account readers may consult Cameron and Schüssler (2023), this collection.

One critical aspect of B-L type dynamos is the spatial dissociation between the source regions of $\mathbf{B_P}$ (on the solar surface) and $\mathbf{B_T}$ (in the deep SCZ). For the B-L dynamo to function effectively, the spatially segregated layers must be connected to complete the dynamo loop. The transport of $\mathbf{B_P}$ generated on the solar surface to the deeper layers of SCZ can occur through various processes. These include meridional circulation (Wang et al. 1991; Choudhuri et al. 1995; Nandy and Choudhuri 2002) which acts as a conveyor belt connecting surface layers to the deep convection zone, turbulent diffusion (Yeates et al. 2008) as well as turbulent pumping Hazra and Nandy (2016). All these processes are an integral part of any flux-transport dynamo model, irrespective of whether the dominant poloidal field generation process is the mean-field or the B-L mechanism.

A new approach towards predictive solar cycle modeling is the coupling of a 2D SFT model to an internal 2D dynamo model – where the output from the first model serves as an upper boundary condition of the second one (Lemerle and Charbonneau 2017). Subsequently, the internal distribution of the toroidal magnetic field (B_T) in the dynamo model generates synthetic sunspots emerging in the SFT model. This model, therefore, has the advantage of incorporating a full non-axisymmetric representation of the solar surface at a much lower numerical cost than the 3D models. The primary weakness of this 2 \times 2D model is its obligation to tackle the different spatial resolutions in the SFT and the dynamo components.

Presenting an elaborate account of all important works on SFT and solar dynamo modelling approaches is beyond the scope of this review; instead we have elaborated only on the primary physical mechanisms that are at the heart of the solar dynamo mechanism. We now turn our focus to processes that are at the basis of solar cycle fluctuations, understanding which is important from the perspective of solar cycle predictions.



4 Physical Processes Influencing Solar Cycle Predictability

As shown in Sect. 2, apart from its about 11-year periodicity, a prominent property of the solar activity record is the strong variability of the cycle amplitudes, including extended intervals of very low activity, e.g., Maunder minimum, or particularly high activity, e.g., modern maximum (Usoskin 2017, Biswas et al. 2023, this collection). Stochastic perturbations inherent in the turbulent solar-stellar convection zones and nonlinearities are two viable candidates for explaining solar cycle fluctuations. Understanding what drives these fluctuations and our ability to account for them either through first principles or observational data assimilation paves the way towards physics-based solar cycle predictions.

4.1 Numerical Weather Forecasts and Nonlinear Time Series Analysis of Solar Activity Proxies

Insights into the development of numerical weather or climate forecasting models over half a century serve as an useful analogy for physics-based solar cycle predictions (Wiin-Nielsen 1991) and could inspire the progress in physics-based solar cycle predictions. Numerical weather forecasts correspond to applying physical laws to the atmosphere, solving mathematical equations associated with these laws, and generating reliable forecasts within a certain timescale. The breakthrough from the 1930s to the 1950s can be classified into two groups. One is the physics of atmospheric dynamics. Vilhelm Bjerknes formulated the atmospheric prediction problem. C. G. Rossby (1939) derived the barotropic vorticity equation and proposed the first theory of the atmospheric long waves, i.e., Rossby waves. J. Charney (1948) developed the quasi-geostrophic theory for calculating the large-scale motions of planetary-scale waves. The second is the genesis of numerical calculation methods and the application of the computational method led by Lewis Fry Richardson and John von Neumann. From 1955 onwards, numerical forecasts generated by computers were issued regularly. People mainly concentrate on four domains to increase the performance of predictions (Kalnay 2003); improve the representation of small-scale physical processes, utilize more comprehensive (spatial and temporal) observational data, use more accurate methods of data assimilation, and utilize more and more powerful supercomputers.

Edward Lorenz (1963) opened the doors of physics based weather forecasting by establishing the importance of nonlinear dynamics in the context of convecting systems and meteorology. In the subsequent remarkable papers (Lorenz 1965, 1969), Lorenz made a fundamental discovery related to the predictability of weather arguing that nonlinearity leads to chaotic dynamics making long-range forecasts impossible. We now know that the chaotic nature of the atmosphere imposes a limit of about two weeks in weather forecasts even with ideal models and perfect observations.

Advances in time-series analysis of non-linear dynamics since the 1980s have made it possible to distinguish between stochastic behavior and deterministic chaos in principle. Strange attractor reconstruction based on correlation integral and embedding dimension (Takens 1981; Grassberger and Procaccia 1983) and the method of surrogate data (Theiler et al. 1992; Paluš and Novotná 1999) are the most widely used methods to look for chaotic behavior. Numerous attempts in this field have been invoked in the literature by analyzing different time series of solar activity proxies, e.g., sunspot number data (Price et al. 1992), sunspot area data (Carbonell et al. 1994), cosmogenic data (Hanslmeier and Brajša 2010), polar faculae (Deng et al. 2016), and so on. However, these studies show highly diverging results. For example, some studies (Mundt et al. 1991; Rozelot 1995; Hanslmeier and Brajša 2010; Deng et al. 2016) report evidence for the presence of low-dimensional deterministic



chaos in solar cycles. On the other hand, others (Price et al. 1992; Carbonell et al. 1994; Mininni et al. 2000) find no evidence that sunspots are generated by a low-dimensional chaotic process. Even in studies showing evidence of chaos, divergent values of the system parameters (e.g., maximum Lyapunov exponent) were estimated – indicating divergent prediction time scales. It is suggested that results claiming the existence of chaos were derived from short scaling regions obtained using very low time delays in the computations for the correlation dimension (Carbonell et al. 1994). Furthermore, the ways to filter or smooth solar activity data also strongly impact the results (Price et al. 1992; Petrovay 2020).

In brief, despite the intensive investigation of solar activity data, there is no consensus on whether chaos or inherent stochastic perturbations or noise, or a combination of these processes drive solar cycle fluctuations. The insufficient length of the solar activity data, that is sparsity of data in phase space, compromises statistical sampling. Clearly distinguishing between stochastic modulation and deterministic chaos remains an outstanding issue. However, driving predictive physical models with observational data of the poloidal component of Sun's magnetic field (which is primarily subject to stochasticity and nonlinear effects) provides a way out of this conundrum.

4.2 Low-Order Models of the Solar Cycle

Model building aims to use our understanding of a physical system by establishing dynamical equations explaining a physical phenomena and furthermore, aid in the interpretation of observational data. One such approach – low-order dynamo models – usually approximate physical processes that occur in a dynamo through truncated equations. Such models have the advantage of exploring a wide variety of solar behavior that is governed by the same underlying mathematical structure, without studying the dynamo process in detail, or making other modeling assumptions. Sections 3.5 and 3.6 of Petrovay (2020) give an overview of this topic classified by two types of low-order models: truncated models and generic normal-form models. See also the review by Lopes et al. (2014). Here we present the progress in this field by classifying them based on different physical processes influencing solar cycle variability and predictability. Although there is no conclusive evidence of the presence or absence of chaos, most recent studies suggest that the irregular component in the variation of solar activity is dominated by stochastic mechanisms.

4.2.1 Deterministic Chaos Subject to Weak Stochastic Perturbations

Such studies assume the non-linear solar dynamo is a chaotic oscillator, subject only to weak stochastic perturbations. The generic normal-form equations are investigated utilizing the theory of non-linear dynamics by bifurcation analysis. The bifurcation sequences are robust. Although the approach has no actual predictive power, they provide an understanding of generic properties and explain the origin of assumed chaotic behavior. Tobias et al. (1995) used a Poincare-Birkhoff normal form for a saddle-node or Hopf bifurcation. Their results show that stellar dynamos are governed by equations that possess the bifurcation structure. Modulation of the basic cycle and chaos are found to be a natural consequence of the successive transitions from a non-magnetic state to periodic cyclic activity and then to periodically modulated cyclic activity followed by chaotically modulated cycles. This behaviour can be further complicated by symmetry-breaking bifurcations that lead to mixed-mode modulation of cyclic behaviour (Knobloch et al. 1998). Trajectories in the phase space spend most of the time in the dipole subspace, displaying modulated cyclic activity, but occasionally flip during a deep grand minimum. Wilmot-Smith et al. (2005) extended the model of Tobias



et al. (1995) to include an axisymmetry-breaking term. The model is able to reproduce some properties found in observations of solar-type stars. Their solution also exhibits clustering of minima, together with periods of reduced and enhanced magnetic activity.

There are also some studies using truncated dynamo models to investigate chaotic behaviour. For example, some argue that mixed modes of symmetry can only appear as a result of symmetry-breaking bifurcations in the nonlinear domain based on a constructed minimal nonlinear α - Ω dynamo (Jennings and Weiss 1991). Tobias (1996) show that grand minima naturally occur in their low-order non-linear α - Ω dynamo if the magnetic Prandtl number is small. The pattern of magnetic activity during grand minima can be contrasted both with sunspot observations and with the cosmogenic record.

Yoshimura (1978) suggested that a time-delay mechanism is intrinsic to the feedback action of a magnetic field on the dynamo process. Wilmot-Smith et al. (2006) constructed a truncated dynamo model to mimic the generation of field components in spatially segregated layers and their communication was mimicked through the use of time delays in a dynamo model involving delay differential equations. A variety of dynamic behaviors including periodic and aperiodic oscillations similar to solar variability arise as a direct consequence of the introduction of time delays in the system. Hazra et al. (2014) extended the model of Wilmot-Smith et al. (2006) by introducing stochastic fluctuations to investigate the solar behaviour during a grand minimum. Recently Tripathi et al. (2021) apply the model with an additive noise to understand the breakdown of stellar gyrochronology relations at about the age of the Sun (van Saders JL et al. 2016). The one-dimensional iterative map is an effective and classical method to investigate the dynamics of a system. Using this method, studies (Durney 2000; Charbonneau 2001) explored the dynamical consequences of the time delay in the B-L type dynamo. As the dynamo number increases beyond criticality, the system exhibits a classical transition to chaos through successive period doubling bifurcations.

4.2.2 Weakly Nonlinear Limit Cycle Affected by Random Noise

Since the non-stationary nature of solar convection is an intimate part of the solar dynamo, a rich body of literature regards that solar variability is largely governed by stochastic perturbations. Random noise has been used to fully mimic the behaviour of the solar cycle (Barnes et al. 1980) while others describe the global behavior of the solar cycle in terms of a Van der Pol oscillator (Mininni et al. 2001); a stochastic parameter corresponding to a stochastic mechanism in the dynamo process was introduced in the Van der Pol equations to model irregularities in the solar cycle were modeled. The mean values and deviations obtained for the periods, rise times, and peak values, were in good agreement with the values obtained from the sunspot time series. Another example is a low-order dynamo model with a stochastic α -effect (Passos and Lopes 2011) in which grand minima episodes manifested; this model is characterized by a non-linear oscillator whose coefficients retain most of the physics behind dynamo theory.

While most low-order models have a loose connection with observations, Cameron and Schüssler (2017) developed a generic normal-form model, whose parameters are all constrained by observations. They introduce multiplicative noise to the generic normal-form model of a weakly nonlinear system near a Hopf bifurcation. Their model reproduces the characteristics of the solar cycle variability on timescales between decades and millennia, including the properties of grand minima, which suggest that the variability of the solar cycle can be understood in terms of a weakly nonlinear limit cycle affected by random noise. In addition, they argue that no intrinsic periodicities apart from the 11-year cycle are required to understand the variability.



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4.3 Babcock-Leighton-Type Kinematic Dynamo Models

Over the past two decades – supported by advances in flux transport models and observational evidence – Babcock-Leighton type solar dynamo models have become the mainstream approach to solar cycle modeling. The B-L mechanism imbibes the processes of emergence of toroidal fields through the convection zone and their subsequent decay and transport by supergranular diffusion and large-scale surface flow fields over the surface, i.e., the SFT processes discussed in Sect. 3.1. Since the SFT processes are directly observed, these act as a source of data assimilation in the B-L dynamo paving the way towards data-driven predictions – akin to what has been achieved in weather models.

4.3.1 Effects of the Meridional Flow and the Time Delay

The meridional flow plays an essential role in the B-L type flux transport dynamo (FTD). The flow strength can modulate not only the cycle strength but also the cycle period. There exists a rich literature describing the effects of the meridional flow on modulation of solar cycles based on FTD models (Yeates et al. 2008; Karak 2010; Nandy et al. 2011; Karak and Choudhuri 2013). Bushby and Tobias (2007) introduced weak stochastic perturbations in the penetration depth of the meridional flow and the results showed significant modulation in the activity cycle; while they argue that this modulation leads to a loss of predictability. Nandy (2021) provides counter arguments pointing out short-term prediction up to one cycle is possible due to the inherent one-cycle memory in the sunspot cycle; see also Yeates et al. (2008), Karak and Nandy (2012), Hazra et al. (2020), Kumar et al. (2021). We note that recently Zhang and Jiang (2022) developed a B-L type dynamo working in the bulk of the convection zone. The model has a much weaker dependence on the flow. Only the flow at the solar surface plays a key role in the polar field generation as in the SFT models implying that the flux transport paradigm is not hostage to meridional circulation being the primary transporter of magnetic flux within the SCZ – as also argued in Hazra and Nandy (2016), Karak and Cameron (2016).

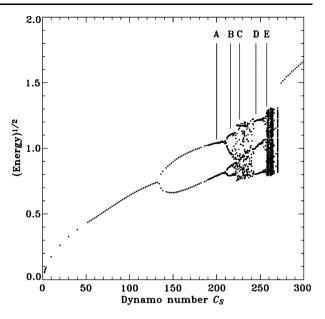
In Sect. 4.2 we have shown numerous attempts at the analysis of the dynamical system using low-order models. For the first time, Charbonneau et al. (2005) presented a series of numerical simulations of the PDE-based 2D B-L type dynamo model incorporating amplitude-limiting quenching nonlinearity. The solutions show a well-defined transition to chaos via a sequence of period-doubling bifurcations as the dynamo numbers $C_S = s_0 R_{\odot}/\eta_t$ (s_0 : strength of the source term, η_t is the turbulent magnetic diffusivity in the Sun's convective envelope) is increased. The results are presented in Fig. 5. Hence they suggest that the time delay inherent to the B-L type dynamo process, acting in conjunction with a simple amplitude-quenching algebraic-type nonlinearity could naturally lead to the observed fluctuations in the amplitude of the solar cycle. The time delay was regarded as the third class of fluctuation mechanisms by Charbonneau et al. (2005). The method was further extended (Charbonneau et al. 2007) to investigate the odd-even pattern in sunspot cycle peak amplitudes. Indeed, it is now being recognized that time delays introduced into the dynamo system due to the finite time necessary for flux transport processes to bridge the source layers of the poloidal and toroidal fields across the convection zone introduces a memory into the system which makes solar cycle predictions a realistic possibility (Nandy 2021).

4.3.2 Observable Nonlinear and Stochastic Mechanisms in the Source Term

Within the framework of the B-L type dynamo, the emergence and decay of tilted bipolar sunspots give rise to the poloidal field. The amount of poloidal field depends on the sunspot



Fig. 5 Bifurcation diagram reconstructed from a sequence of numerical 2D B-L type dynamo solutions with increasing dynamo numbers C_S . Vertical lines labeled 'A' to 'E' correspond to representative 2-periodic, 4-periodic, 5-periodic, 3-periodic, and chaotic solutions, respectively. Image reproduced with permission from Charbonneau et al. (2005), copyright by the American Astronomical Society



properties, e.g., the tilt angle of the bipolar sunspots, which show both the systematic property resulting from Joy's law and the stochastic one due to the tilt scatter. Hence, the B-L mechanism has an inherent randomness. Studies like Charbonneau and Dikpati (2000), Passos et al. (2014), Kitchatinov et al. (2018), Hazra and Nandy (2019a), Saha et al. (2022) took the stochastic fluctuation in the poloidal field source term as a free parameter and investigated their possible effects on the cycle modulation. Based on a B-L type solar dynamo with an additional mean-field α -effect, Sanchez et al. (2014) quantified the intrinsic limit of predictability, i.e., e-folding time τ , which is the equivalent of the two weeks for the weather forecast. As expected, the e-folding time is shown to decrease corresponding to a short forecast horizon, with the increase of the α -effect.

Studies (Kitchatinov and Olemskoy 2011; Choudhuri and Karak 2012; Olemskoy et al. 2013; Karak and Miesch 2017) attempted to estimate the parameters of the B-L mechanism and their fluctuations using historical sunspot data. Jiang et al. (2014a) measured the tiltangle scatter using the observed tilt-angle data and quantified the effects of this scatter on the evolution of the solar surface field using SFT simulations with flux input based upon the recorded sunspot groups. The result showed that the effect of the scatter on the evolution of the large-scale magnetic field at the solar surface reaches a level of over 30%. When a BMR with area A and tilt angle α emerges, it has the (initial) axial dipole field strength $D_i \propto A^{1.5} \sin \alpha$. We define the final contribution of a BMR to the axial dipole field as the final axial dipole field strength D_f . Jiang et al. (2014a) show that D_f has the Gaussian latitudinal dependence. The result was confirmed by others (Nagy et al. 2017; Whitbread et al. 2018; Petrovay et al. 2020b), that is

$$D_f = D_i \exp(-\lambda^2/\lambda_R^2), \tag{5}$$

where λ_R is determined by the ratio of equatorial flow divergence to diffusivity. Wang et al. (2021) further generalized the result to ARs with realistic configuration, which usually show large differences in evolution from the idealized BMR approximation for δ -type ARs (Jiang et al. 2019; Yeates 2020). Hence big ARs emerging close to the equator could have big



effects on the polar field evolution, and on the subsequent cycle evolution based on the correlation between the polar field at cycle minimum and the subsequent cycle strength. These are referred to as rogue ARs by Nagy et al. (2017). Jiang et al. (2015) demonstrated that these low-latitude regions with abnormal polarity could indeed be the cause of the weak polar field at the end of cycle 23, hence the low amplitude of cycle 24. Simulations by Nagy et al. (2017) indicate that in the most extreme case, such an event could lead to a grand minimum; they argue that the emergence of rogue ARs in the late phase of a cycle may limit the scope of predicting the dipole moment (and polar field amplitude) at the minimum of a cycle. However, it is likely that the impact of such rogue regions may be estimated through the ensemble prediction approach (Cameron et al. 2016; Bhowmik and Nandy 2018; Jiang et al. 2018).

The stochastic properties of the BMRs emergence mentioned above provide the observable stochastic mechanisms in solar cycle modulation. The systematic properties of the BMRs emergence have recently been suggested to be observable nonlinearities. Historical data show that the cycle amplitude has an anti-correlation with the mean tilt angle of BMRs (Dasi-Espuig et al. 2010; Jiao et al. 2021) and a positive correlation with the mean emergence latitudes (Solanki et al. 2008; Jiang et al. 2011). Jiang (2020) investigated the effects of the latitude and tilt's properties on the solar cycle, which are referred to as latitudinal quenching and tilt quenching, respectively as depicted in Fig. 6 (also see Jha et al. 2020). They defined the final total dipole moment, which is the total dipole moment generated by all sunspot emergence during a whole cycle. For a stronger cycle, more ARs emerge at higher latitudes, limiting the amount of flux cancellation between the leading spots of opposite polarities through cross-equatorial interactions. Effectively, this causes a weaker impact on the dipole moment evolution as magnetic flux from both leading and following spots of opposite polarities advect towards the poles. Additionally, ARs from stronger cycles are likely to have smaller mean tilt angles, thus collectively contributing less to the dipole moment. In comparison, in a weaker solar cycle, more ARs emerging at lower latitudes ensure enhanced cross-equatorial flux cancellation, thereby, a larger effect on the dipole moment. Weaker cycles also tend to have ARs with higher mean tilt angles, which again will be more impactful in determining the dipole moment evolution. Thus, both forms of quenching lead to the expected final total dipolar moment being enhanced for weak cycles and saturated to a nearly constant value for normal and strong cycles. This could be an explanation of the observed long-term solar cycle variability, e.g., the odd-even rule. Karak (2020) verified that latitudinal quenching is a potential mechanism for limiting the magnetic field growth in the Sun using a three-dimensional B-L type dynamo model. Talafha et al. (2022) systematically explored the relative importance played by these two forms of quenching in the solar dynamo showing that this is governed by λ_R .

5 Physics-Based Solar Cycle Predictions

The importance of the dipole moment (or the average polar field) in solar cycle predictions is established through observation and dynamo theory. The most successful empirical method for solar cycle predictions based on the polar field precursor (Schatten et al. 1978) in fact predated the solar dynamo model based predictions. Thus, physics-based predictions of the solar cycle, in general, are either based on SFT simulations aiming to estimate the dipole moment (related to polar flux) at the cycle minimum or involve dynamo simulations with modified poloidal field source term according to the observed (or simulated) dipole moment at the cycle minimum. Nandy (2002) first alluded to the possibility of developing data-driven



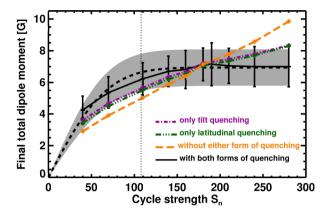


Fig. 6 Effects of observable nonlinear and stochastic mechanisms in the source term on the poloidal field generation. The black solid curve indicates the expected values from 100 SFT simulations using random sunspot group realizations including latitudinal and tilt quenching. Error bars correspond to the 1σ standard deviation, caused by the randomness in the properties of sunspot groups. Green dashed—triple-dotted and purple dashed—dotted curves show the expected values for SFT simulations with only the latitudinal and tilt quenching, respectively. The orange dashed curve shows the expected value of SFT simulations without latitudinal or tilt quenching. Image reproduced with permission from Jiang (2020), copyright by the American Astronomical Society

predictive solar dynamo models by utilizing the observed poloidal field as inputs; although this particular branch of solar cycle predictions is relatively new, significant progress has been achieved through contributions from multiple works predicting the past cycle 24 and present cycle 25 using physics-based models.

5.1 Role of SFT Models in Solar Cycle Predictions

Despite the dissimilarities among different SFT models regarding their treatments of the emerged sunspot statistics and observed transport processes on the photosphere, SFT models have played a major role in physics-based solar cycle predictions, especially of cycle 25. The idea lies in the fact that the Sun's magnetic axial dipole moment at the end of a solar cycle is strongly correlated with the following cycle's peak amplitude. This positive correlation is found from observation spanning multiple cycles (Muñoz-Jaramillo et al. 2013) and is also supported by the principles of the solar dynamo mechanism (see Sect. 3.2 for more details).

Using SFT simulations, Cameron et al. (2016) presented the first prediction of cycle 25 about four years before the cycle minimum (which occurred at the end of 2019). Their simulation started with the observed synoptic magnetogram at the end of 2015. The sunspot emergence statistics in the declining phase (years: 2016–2020) of cycle 24 were generated using 50 randomizations which included uncertainties associated with the emergence timing, latitude-longitude position, tilt angle and magnetic flux of the sunspots. They provided a possible range of the axial dipole moment at the end of cycle 24. Based on the positive correlation between dipole moment and the following cycle amplitude, they predicted a moderately strong sunspot cycle 25. Hathaway and Upton (2016), Upton and Hathaway (2018) took a similar approach to estimate the axial dipole moment at the end of cycle 24 using their AFT simulations. However, the sunspot statistics corresponding to the declining phase of cycle 24 were taken from the declining phase of solar cycle 14. The uncertainties in predicting the dipole moment were realised by considering stochastic variations in the convective motion details, sunspot tilt angles, and changes in the meridional flow profile. Their



predicted dipole moment amplitude suggested that cycle 25 would be a weak to moderate cycle. Iijima et al. (2017) argued that the axial dipole moment does not vary significantly in the last three years of the declining phase of any sunspot cycle. Thus to predict the dipole moment at the end of cycle 24, they initiated their SFT simulation in 2017 with an observed synoptic magnetogram and continued till the end of 2019 without assimilating any sunspots. Their prediction suggested a weak solar cycle 25. The importance of correctly simulating the surface magnetic field distribution and their consecutive inclusion in an interior dynamo model was extensively utilized by other studies (Labonville et al. 2019; Bhowmik and Nandy 2018) for cycle predictions.

As addressed in previous sections, the randomness in sunspot emergence properties plays an important role in the final amplitude of the axial dipole moment at cycle minimum. Assimilating the uncertainties associated with the sunspots and the observed magnetogram used as the initial condition in the SFT simulations, Jiang et al. (2018) demonstrated a generalized scheme to investigate the predictability of the solar cycle over one cycle. The scheme requires at least three years of sunspot observation for the ongoing cycle to predict the statistical properties of the sunspots during the remaining phase. Their prediction of cycle 25 based on this scheme suggests a weak-to-moderate cycle peak.

5.2 Dynamo-Based Solar Cycle Predictions with Data Assimilation

The B-L type 2D flux transport dynamo models were utilized for the first time to predict cycle 24 during the mid-2000s (Dikpati et al. 2006; Choudhuri et al. 2007). However, the only two physics-based predictions of cycle 24 diverged significantly from each other (with a difference of ~ 100 sunspots during the maximum). Despite using similar dynamo models, such divergence can arise from two aspects: differences in the dominating flux transport processes (Yeates et al. 2008) and how the B-L source term is designed according to the observed magnetic field. Exploring these two points is crucial for understanding the physics behind sunspot cycle predictions as well as providing realistic forecasts.

All kinematic flux transport dynamo models consider the following transport processes at the least: differential rotation, meridional circulation and magnetic diffusion. They use analytical functions corresponding to the observed differential rotation (Howe 2009). The observed meridional flow on the solar surface sets constraints for the meridional circulation profile used within the SCZ. The exact structuring of this flow at different layers of the SCZ still requires further observational support (Hanasoge 2022), but recent helioseismic studies suggest a one-cell meridional circulation (Rajaguru and Antia 2015; Liang et al. 2018). Nonetheless, these models assume an equatorward branch of the meridional flow near the tachocline which ensures observed latitudinal propagation of sunspot activity belts in both hemispheres and an appropriate cycle duration (Choudhuri et al. 1995; Dikpati and Charbonneau 1999; Nandy and Choudhuri 2002; Chatterjee et al. 2004; Hazra et al. 2014; Hazra and Nandy 2019b). Furthermore, the amplitude and profiles of magnetic diffusivity are also based on analytical functions, which only vary with the depth of the SCZ (Muñoz-Jaramillo et al. 2011).

However, based on the strength of diffusivity, flux transport dynamo models behave differently and can be categorized into two major classes: advection dominated (diffusivity order, $\eta \sim 10^{10}$ cm² s⁻¹, see Dikpati and Charbonneau 1999) and diffusion dominated ($\eta \sim 10^{12}$ cm² s⁻¹, see Choudhuri et al. 2007). The strength of the diffusivity decides which transport mechanism between meridional circulation and magnetic diffusivity will be more effective for convecting $\mathbf{B_P}$ to the deeper layers of the SCZ (Yeates et al. 2008). It also determines whether $\mathbf{B_P}$ associated with multiple past solar cycles can survive in the SCZ at



the prescribed diffusivity and contribute simultaneously to the generation of new B_T of the following cycle (Yeates et al. 2008; Jiang et al. 2007). However, the inclusion of turbulent pumping (Karak and Nandy 2012; Hazra and Nandy 2016) as an additional transport process in flux transport dynamo diminishes the difference between the advection-dominated and diffusion-dominated regimes. All these results are crucial for estimating the dynamical memory of the solar dynamo models and their ability to accurately predict the future solar cycle amplitude. Dynamical memory is a measure of determining the range of temporal association of the poloidal field (B_P) of a certain sunspot cycle (say, n^{th}) with the toroidal field (B_T) of following cycles (say, $n+1^{th}$, $n+2^{th}$, $n+3^{th}$, etc.). Note that for advection-dominated dynamo models, the dynamical memory is about two solar cycles, whereas it's about half a solar cycle for diffusion-dominated dynamo models or in models with turbulent pumping (Yeates et al. 2008; Karak and Nandy 2012).

Besides the transport parameters, how we assimilate observational data to model the poloidal field (B_P) source will influence the successive generation of the toroidal field (B_T) , thus is crucial for solar cycle predictions. Below, we discuss this aspect of 'data-driven' dynamo models in the context of solar cycle prediction.

As mentioned in Sect. 3.2, surface flux transport processes acting on emerged active regions produces the large-scale photospheric field and serves as a reliable means for poloidal field generation. Multiple efforts have been made to assimilate the observed surface data in flux transport dynamo models. Dikpati and Gilman (2006) included observed sunspot group areas to formulate the B_P source while assuming all spots of any solar cycle are distributed in the same latitudinal belt (between 5° and 35°) and have similar tilt angles. However, their data-driven model failed to correctly predict the solar cycle 24 peak (Dikpati et al. 2006). The primary reasons for this disparity were that the idealized realization of the sunspots (fixed latitude and tilt angle) results in a poloidal source at the minimum (of n^{th} cycle) directly proportional to the preceding cycle (n^{th}) amplitude and that the low magnetic diffusivity in their flux transport dynamo model increased the dynamical memory to more than two solar cycles. Thus according to their model, not only does a strong solar cycle (n^{th}) produce a strong poloidal source, its strength influences several following solar cycles ($n+1^{th}$, $n+2^{th}$ and $n+3^{th}$).

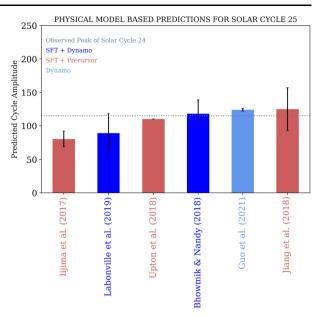
In contrast, Choudhuri et al. (2007) and Jiang et al. (2007) used a diffusion-dominated flux transport dynamo model to predict sunspot cycle 24. For modeling the B_P source, they relied on observed large-scale surface magnetic field distribution (for example, the axial dipole moment) during the solar cycle 23 minimum. Their prediction was a good match to the observed peak of cycle 24. The positive correlation between $B_P(n)$ and the $B_T(n+1)$ using a diffusion-dominated dynamo model where the dynamic memory is half a solar cycle ensured the success of their prediction. Guo et al. (2021) took a similar approach by combining observed axial dipole moment to a diffusion-dominated dynamo model to predict cycle 25. In a recent review, Nandy (2021) discusses in details how observations and stochastically forced dynamo simulations support only a short half- to one-cycle memory in the solar cycle, suggesting that the latter class of dynamo models are the right approach to take for solar cycle predictions.

Recently, Bhowmik and Nandy (2018) assimilated a series of surface magnetic field distribution at eight successive solar minima (cycles 16-23 minima) in a flux transport dynamo model (diffusion-dominated). The surface maps were obtained from their calibrated century-scale SFT simulation, which assimilates the observed statistics of emerging bipolar sunspot pairs during that period. Their coupled SFT-dynamo simulations reproduced past solar cycles (17-23) with reasonable accuracy (except cycle 19). The same methodology was utilized to provide an ensemble forecast for cycle 25 while assimilating the predicted



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Fig. 7 Comparison of physical model-based predictions of solar cycle 25 peak amplitude. The average of the six predictions is 107.75 sunspots (with $\pm 1\sigma = 17.15$). Details on each of these predictions are described in Sects. 5.1 and 5.2



surface field distributions at cycle 24 minimum from their SFT simulations. Their prediction indicates a weak cycle 25 (with a peak SSN of 118 and a range from 109 – 139), similar to, or slightly stronger than cycle 24. Note that the upper bound of the Bhowmik and Nandy (2018) prediction as reported in Nandy (2021) was misreported as 155 but should have been 139. The 2 × 2D SFT-dynamo model by Lemerle and Charbonneau (2017) is another example of coupling the surface magnetic field to the internal dynamo, which occurs more intimately in their model. Labonville et al. (2019) utilized the same model to assimilate the series of BMRs observed during cycles 23 and 24 (from Yeates et al. 2007) into the SFT part of the simulations. They first calibrated the model by including cycle 23 data to produce an ensemble forecast for cycle 24 and subsequently comparing it with observation. They then assimilated the BMRs series for cycles 23 and 24 to present an ensemble forecast for cycle 25, including its amplitude (weaker than cycle 24), rising and declining phases and northern and southern asymmetry.

All physics-based predictions for cycle 25 are depicted in Fig. 7.

5.3 Comparison with Other Prediction Methods

The methodologies of predictions are not restricted to physical modeling only. They can be based on (a) precursor technique-based forecasts, (b) non-linear mathematical modeling, (c) statistical forecasts, (d) machine learning and neural network, (e) spectral methods etc. (Petrovay 2020; Nandy 2021). Note that most of the precursor-based forecasts consider the physics of solar dynamo and cycle predictability to some extent without performing any computational modeling. For example, based on semi-empirical and semi-physical approaches Hazra and Choudhuri (2019) and Kumar et al. (2022) predicted solar cycle 25 amplitude, where the latter group claimed that polar field evolution after the polarity reversal exhibits similar features like the Waldmeier effect and can be utilized for cycle predictions. Nonetheless, the forecasts based on diverging techniques show a wide variation in predicted cycle amplitudes: with peak sunspot numbers ranging between 50 and 250 for solar cycle



25 (see Fig. 3 of Nandy 2021). In that regard, physics-based predictions of cycle 25 have reached a consensus with an average of 107.75 sunspots (with $\pm 1\sigma = 17.15$). In contrast, for cycle 24 physics-based predictions, the difference between the two predicted peaks was more than 100 sunspots.

6 Summary and Future Prospects

It is noteworthy that while the only two physics-based predictions of solar cycle 24 diverged significantly from each other, physics-based predictions of cycle 25 show significantly more convergence. This is possibly an indication of increasing understanding of the solar dynamo process – as argued in Nandy (2002) – and advances in assimilating observational data in the computational models used for predictions. However, there are significant improvements that are still necessary in the adopted modeling approaches.

While all physical models of solar cycle predictions have been 2D in nature, global 3D dynamo models have the promise of capturing the surface flux transport dynamics and internal magnetic field evolution self-consistently. Some recent works have solved the induction equation in three spatial dimensions within a dynamo framework thus going beyond the 2D axisymmetric models (Yeates and Muñoz-Jaramillo 2013; Miesch and Dikpati 2014; Hazra et al. 2017; Karak and Miesch 2017; Kumar et al. 2019; Whitbread et al. 2019). These are B-L type dynamos working in a kinematic mode with modules to incorporate realistic sunspot emergence and decay of the associated flux. These models provide the opportunity of further development towards dynamical models, imbibing in-built axisymmetric and nonaxisymmetric feedback mechanisms (Nagy et al. 2020), thus slowly closing the gap between these phenomenological B-L type dynamo models and the full MHD dynamo models. However, such models still have two significant discrepancies: the polar field generated on the surface is much stronger than the observed order of magnitude, and sunspot emergence at higher latitudes is prevented artificially. Nonetheless, these models hold the promise of selfconsistently imbibing elements of the surface flux transport dynamics leading to polar field reversal and build-up as well as solar internal magnetic field evolution, which 2D models cannot.

Finally, we end by posing the following provocative questions. From a purely utilitarian view, are dynamo models at all necessary for solar cycle predictions and can they provide higher accuracy or advantage compared to empirical polar field precursor based techniques? There is no doubt that physical approaches based on surface flux transport models and dynamo models have significantly advanced our understanding of solar cycle predictability; however, the resolution of the above questions are fundamental to sustained growth prospects of this field of research.

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Author Contribution D.N. coordinated and planned the writing of this review in consultation with all authors. P.B. managed and led the manuscript writing process to which all authors contributed.

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Declarations

Competing Interests The authors declare no conflicts of interest.

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Authors and Affiliations

Prantika Bhowmik¹ · Jie Jiang^{2,3} · Lisa Upton⁴ · Alexandre Lemerle^{5,6} · Dibyendu Nandy^{7,8}

P. Bhowmik

prantika.bhowmik@durham.ac.uk

J. Jiang

jiejiang@buaa.edu.cn

L. Upton

lisa.upton@swri.org

A. Lemerle

alexandre.lemerle@umontreal.ca



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D. Nandy dnandi@iiserkol.ac.in

Department of Mathematical Sciences, Durham University, Stockton Road, Durham, DH1 3LE, UK

- ² School of Space and Environment, Beihang University, Beijing, China
- ³ Key Laboratory of Space Environment Monitoring and Information Processing of MIIT, Beijing, China
- Department of Solar and Heliospheric Physics, Southwest Research Institute, Boulder, 80302, CO, USA
- Département de Physique, Collège de Bois-de-Boulogne, Montréal, Québec, Canada
- ⁶ Département de Physique, Université de Montréal, Montréal, Québec, Canada
- Center of Excellence in Space Sciences India, Indian Institute of Science Education and Research Kolkata, Mohanpur, 741246, West Bengal, India
- Department of Physical Sciences, Indian Institute of Science Education and Research Kolkata, Mohanpur, 741246, West Bengal, India

