The aim of this study is to demonstrate the effect of particle size on semiconductor properties; artificial intelligence is being used for the research methods. As a result, we picked cadmium sulfide (CdS), which is a unique semiconductor material that is employed in a broad variety of current applications. Given that CdS has distinct electrical and optical characteristics, it may be employed in the production of solar cells, for example. Solar cells, as is also well known, have become an essential source of energy in the world. Within the visible range (500-700 nm), we create one layer of bulk CdS and one layer of nano-CdS air bulk CdS air and air nano-CdS air. We used a number of instrumentation methods to investigate the naked CdS nanoparticles, including XRD, SEM-EDX, UV-Vis spectroscopy, TEM, XPS, and PL spectroscopy, among others. The results show that for bulk CdS at normal incidence, the transmittance is $T = 45$, and for nano-CdS with particle size 3 nm, the transmittance is $T = 85.8$, with transverse-electric (S-polarized) and transverse-magnetic (P-polarized) transmittances of $TE = 75$ and $TM = 80$, respectively.
1. Introduction

The photodiode capabilities of CdS single nanoribbons were fully examined, including spectrum response, light intensity response, and temporal response. Traditional film and bulk CdS materials have a far slower response speed than CdS nanoribbons, and nanoribbon size has a significant influence on response speed, with smaller CdS nanoribbons having a higher reaction speed. The enormous surface-to-volume ratio and high single-crystal quality of CdS nanoribbons, as well as the reduction in the recombination barrier in nanostructures, account for the high photosensitivity and light response speed. The absorption of ambient gas (mostly oxygen) can significantly affect the photosensitivity of CdS nanoribbons by trapping electrons from the nanoribbons, according to findings in a different environment.

Solar cells began to bring life to homes and businesses in both urban and rural areas. Solar cell power systems have become one of the world’s most important sources of energy [1].

At this time, we cannot say that the adoption of photovoltaic (PV) systems is the most effective way of addressing global energy consumption issues. Renewable energy resources, on the other hand, are a viable alternative for sustaining and supporting a country’s economy [2]. P-n junction building blocks are used to generate energy in solar cells. They are made of several semiconductor materials that absorb different wavelengths of sunlight.

The influence of particle size on the properties of CdS has been studied. This study investigates nanotin films because of their unique electrical and optical characteristics. Cadmium sulfide (CdS) is an important semiconductor that may be used in a variety of applications. CdS is a kind of cadmium supplied that is researched in this paper [3, 4].

Solar cell usage began in both urban and rural areas. Currently, we cannot say that photovoltaic (PV) systems are the most effective way to decrease global energy demand. However, renewable energy sources may improve a country’s economy. P-n junctions produce energy in solar cells. Their absorption of light varies. Particle size affects CdS characteristics. Cadmium sulfide (CdS) is a versatile semiconductor with unique electrical and visual properties.

2. Reflectance and Transmittance

Nanoscience and nanotechnology basically deal with structure, characterization, exploration, and utilization of nanostructured materials. Nanostructures establish an intermediate between the molecular scale and infinite bulk. Individual nanostructures comprise bunches, quantum dots, nanocrystals, nanowires, and nanotubes [4–7].

Calculating optical absorption spectra of thin films allows us to calculate the Urach energy (Eα), optical energy gap (Eg), absorption coefficient, and nature of the transition. The optical absorption coefficient α can be calculated by using the thickness (t) by estimating T(l) and R(l) as follows [8]:

$$\alpha = \frac{1}{l} \ln \left( \frac{1 - R^2}{T} \right).$$  (1)

The photon energy (hv) and the absorption coefficient α are related by the equation [9]

$$\alpha hv = \beta (hv - E_g)^n.$$  (2)

The band tailing parameter (a constant) is determined by the effective mass of electrons and holes as well as the optical middle density. The nature of the transition determines the power exponent [8].

The Effective Mass Approximation (EMA) or Brus model is the most popular to clarify how the value of the energy band gap depends on the semiconductor quantum dot size. Figure 1 shows that this model depends on the value of the effective mass m* and m* of electrons and holes, according to the Brus equation [10–12]:

As seen in Figure 1, the Fresnel amplitude reflection coefficient (r) at a point of interaction between two nonabsorbing media at normal incidence may be calculated using the equation $r = (n_1 - n_0)(n_1 + n_0)$. $n_1$ and $n_2$ are the (actual) refraction directories of the two media [12]. The calculations for transparent media are made using the formulas below.

$$T = \frac{y_2}{y_1} = \frac{4y_1y_2}{(y_1 + y_2)^2},$$  (3)

where $\tau$ is the transmission coefficient $r = 2y_2/[y_2 + y_1]$. For an oblique incidence, the reflectivity and the transmittance are known as the relationship

$$R = \left( \frac{n_0 - n_1}{n_0 + n_1} \right)^2,$$  (4)

$$T = \frac{4n_0 \Re (n_1)}{(n_0 + n_1)(n_0 - n_1)}$$  (5)

where $n_\rho = y_\rho / \cos \nu$ and $n_\iota = y_\iota \cos \nu$ and $\nu$ is the angle of incidence [13]. Nanoscience and nanotechnology basically deal with structure, characterization, exploration, and utilization of nanostructured materials. Nanostructures establish an intermediate between the molecular scale and infinite bulk. Individual nanostructures comprise bunches, quantum dots, nanocrystals, nanowires, and nanotubes [14]. We can determine the Urach energy (Eα), optical energy gap (Eg), absorption coefficient, and nature of the transition by calculating optical absorption spectra of thin films. One could use the thickness (t) to determine the optical absorption coefficient α by estimating T(l) and R(l) as follows [15]:

$$\alpha = \frac{1}{l} \ln \left( \frac{1 - R^2}{T} \right).$$  (6)

The photon energy (hv) and the absorption coefficient α are related by the equation [16]

$$\alpha hv = \beta (hv - E_g)^n.$$  (7)
where $\beta$ is the band tailing parameter (a constant), which depends on the effective mass of the electrons and holes and the optical middle density.

3. Results and Discussion

The nature of the transition determines the value of the power exponent $n$. The Effective Mass Approximation (EMA), often known as the Brus model, is the most extensively used method for explaining how the energy band gap fluctuates with semiconductor quantum dot size in semiconductor devices. According to the Brus equation, this model is based on the effective masses of electrons and holes, $m_e^*$ and $m_h^*$, respectively. A computer program (MATLAB) was built to determine CdS transmittance at visible wavelengths (500-700 nm), as well as the influence of nanoparticle sizes on transmittance and other parameters such as the energy gap and refractive index.

The Effective Mass Approximation (EMA) equation, the radius of the quantum dot, and the effective band gap are all used in this application. The values of the physical attributes employed in the software are shown in Table 1.

The coating is made up of CdS and an air substrate with a refractive index of $n = 1$ and a wavelength transmittance of 450 nm. Figure 2 shows the architecture of deep learning of CdS, which is $T = 53$, at a wavelength of 450 nanometers.

The figure shows the transmittance for bulk CdS and incidence angles of 27°, 45°, and 9°. Table 2 shows the transmittance values of a 450 nm wavelength at oblique incidence based on optical thicknesses of layer that was quartered-wavelength long and refractive index and angle of incidence light using equation (4), and Table 2 shows the transmittance values of a 450 nm wavelength at oblique incidence based on destructive and constructive.

Semiconductor quantum dots (QDs) have gotten a lot of attention because of their unusual size-dependent optical and electrical properties [16]. In optoelectronic devices, these different features have a wide range of applications [17]. CdS, CdSe, and CdTe are examples of visible light-emitting group II-VI semiconductor nanocrystals [1].

Cadmium sulfide (CdS) is a semiconductor that has an $E_g$ eV straight band gap. There are uses for photodetectors, optoelectronics, and solar cells [18]. A variety of sulfur sources have been investigated for the production of CdS nanoparticles. The most often used solution-based preparations of man-sized CdS are sodium-supplied hydrogen-supplied gas [7] and theorem [9]. As a result, much effort has been put into creating high-quality CdS nanocrystal manufacturing methods such as controlled precipitation, high-temperature hemolysis of single precursors, and quick hot-injection-based synthesis. The nucleation and growth stages are effectively separated by the quick hot-injection-based synthesis. To control the size of semiconductor nanocrystallites, a number of preparation procedures have been developed, including (a) using stabilizers and polymers.

(a) Exclusion chromatography for size fractionation

(b) Chromatography for size selection

CdS nanocrystals exhibit green, yellow, and red bands in their photoluminescence (PL) spectra [19]. Band-edge emissions are connected with the green and yellow bands, whereas surface imperfections or surface states are associated with the low-energy band or red band [20]. When nanocrystals are not molecularly contained (naked), the red band typically governs the PL spectra [21]. When the broad band is between 700 and 800 nm, complex defects such as cadmium vacancies [22] or sulfur vacancies may be to blame.
CdS is dimorphic in its cubic (zinc mix type) and hexagonal (wurtzite type) forms [23]. The optical properties of semiconductor nanocrystal particles are determined by their crystalline structure. The effective masses of electrons and holes in related electronic bands will change considerably due to crystalline phase fluctuations. Unlike the cubic phase, which is kinetically controlled (metastable), the hexagonal phase of CdS is thermodynamically regulated (stable) [24]. As a result, the researchers developed a technique for producing thermodynamically stable nanoscale CdS materials. The goal of this research is to see how the mass concentration of Cd2+ ions and the Cd2+ ion source impact the crystalline phase of bare CdS nanoparticles. Based on the hemolysis of a single-source organometallic precursor, we created a simple, low-cost, and environmentally friendly synthesis strategy in this study. Tong and Zhu have been together for quite some time. To study the naked CdS nanoparticles, we employed a variety of instrumentation techniques, including XRD, SEM-EDX, UV-Vis spectroscopy, TEM, XPS, and PL spectroscopy.

The transmittance is also affected by the refractive index of the nanoparticles. The transmittance increases as the refractive index of the nanoparticles lowers, as seen in Figure 2. Inversely, as illustrated in figures, as the energy gap of the effective band gap increases, the transmittance increases.

When particle size ($P = 2R$) is equal to or less than the Bohr exciton radius ($\sigma = 3$ nm), “the effect of quantum confinement” arises and “the effect of quantum confinement” increases dramatically, leading to improved material transmittance.

4. Conclusion

Thus, particle size affects several physical parameters of cadmium sulfide. When the particle radius is equal to or less than the excitation’s Bohr radius, due to quantum confinement, the energy gap increases and the index of refraction narrows with decreasing particle size.

Particle size impacts the nanoenergy gap and refractive index (Figure 2). The nanoenergy gap widens as particle size shrinks and refractive index falls. As seen in Figure 1, changing characteristics affect transmittance. The optimum coating for solar cells has a transmittance of 73.9 for 3 nm particles and $TE = 71$ and $TM = 78$ for nonnormal incidence = $32^\circ$ at 450 nm. In the future, a new material with a different nanothickness may be added to the first layer to improve transmittance.

Data Availability

The data used to support the findings of this study are included within the article.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

Authors’ Contributions

The authors confirm contribution to the paper as follows: (i) study conception and design: Walid Theib Mohammad, Dr. Sherin Hassan Mabrouk, and Dr. Rania Mohammed Abd Elgawad Mostafa; (ii) data collection: Ahmad Maher Al Sayeh, Malik Bader Alazzam, and Md Kawser Ahmed; (iii) analysis and interpretation of results: Dr. Mona Abdelmo- neim Abdelmabood Ebrahim, Samar Zuhair Alshawwa, Dr. Heba Abdelazeem hassen Ismail, and Dr. Manal Mahrous Abdulhamed Mohamed; (iv) draft manuscript preparation: Mohammad Bani Younis and Ms. Sara wans Alshmmry. All authors reviewed the results and approved the final version of the manuscript.
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